

How you look at a picture determines if you will remember it



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Dataset: <http://figrim.mit.edu>
Paper: Bylinskii, Z., Isola, P., Bainbridge, C., Torralba, A., Oliva, A. "Intrinsic and Extrinsic Effects on Image Memorability", Vision Research 2015.

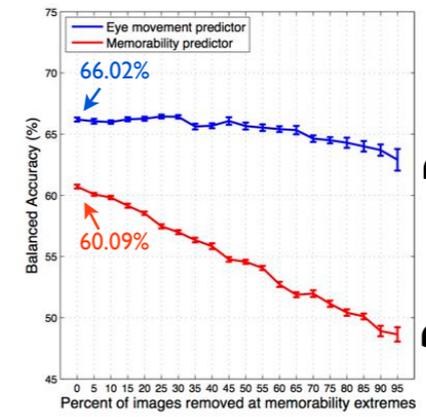
Abstract: We introduce an approach to predict user memory of images: specifically, to infer if a particular user will remember a particular image at a later time. We measure an unconscious motor signature of memory: eye fixations while a user views images. We train a boosting classifier to differentiate eye movements that lead to a successful memory of an image from those that lead the image to be forgotten.

goal: predict whether user will remember (recognize) this image later in the sequence



we use eye-movements at encoding to make predictions
in-lab eyetracking experiments with:
- 630 target natural scene images
- 42 participants (~16.2 per image)

$$\text{balanced accuracy} = \frac{0.5 \times \text{true positives}}{\text{true positives} + \text{false negatives}} + \frac{0.5 \times \text{true negatives}}{\text{true negatives} + \text{false positives}}$$



A crowd can be used to make significantly above-chance predictions for an individual user.



When images at the memorability extremes are removed, eye-movements remain robust predictors for an individual user.

1 coarse binning and smoothing

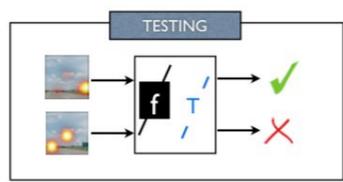
We compute a continuous **fixation map** by coarse binning and smoothing fixation locations (20x20 grid, sigma=2)

2 learn classifier for successful fixations on this image vs successful fixations on other images

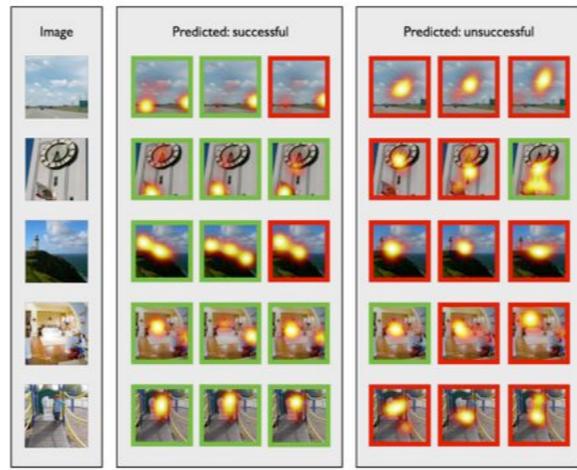
Just the task of classifying whether fixations come from this VS other image achieves 79.7% **balanced accuracy**.

3 find threshold for separating successful vs unsuccessful fixations on this image

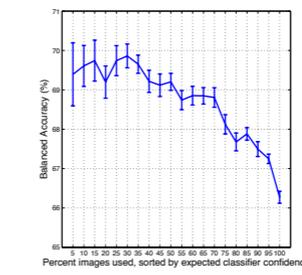
We learn a threshold under each image's **exemplar classifier** for differentiating successful from unsuccessful fixation maps on the image.



A set of fixations that are more likely to have been elicited by the current image as opposed to some other image are **predicted successful** encoding fixations that will lead to a correct recognition of the image.



Not all image are equally predictable (successful should be distinguishable from unsuccessful fixations)



Applications: an automatic system that monitors the eye movements of a user to determine when to re-present visual material until there is confidence that the user has properly encoded the content.

For more robust performance, can select more predictable images for a given application.