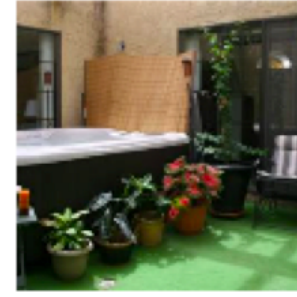


Computational Visual Cognition: Predicting Image Memorability



Aude Oliva

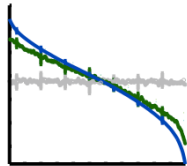
**Computer Science and Artificial Intelligence Lab (CSAIL)
Massachusetts Institute of Technology
oliva@csail.mit.edu**



Predicting Image Memorability

What could we do with such knowledge and technology?

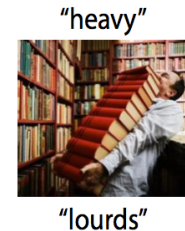
**Understand
human
memory**



**Diagnose
memory
problems**



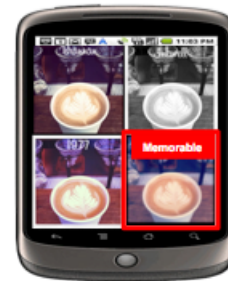
**Design
mnemonic
aids**



Advertising



**Mobile
Applications**



**Retrieve
better images
from search**



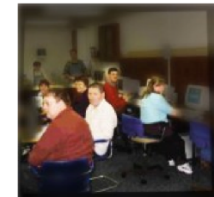
Logos



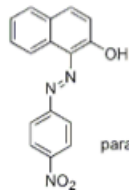
**Social
Networking**



**Make an
image more
memorable**



Education



**Face
Memorability**



**Summarize
photo album
or video**



Nature of visual long term representations

What we know ...

Standing (1973)

10,000 images

83% Recognition

*... people can
remember thousands
of images*

What we don't know ...

*... what people are remembering
for each item?*



According to Standing

“Basically, my recollection is that we just separated the pictures into **distinct thematic categories**: e.g. cars, animals, single-person, 2-people, plants, etc.) Only a few slides were selected which fell into each category, and they were visually distinct.”



“Abstract Only



Sparse Details



Highly Detailed

Massive Memory Experiment I

A stream of objects will be presented on the screen for ~ 3 second each.

Your primary task:

Remember them ALL!

afterwards you will be tested with...

Completely different objects...



Different exemplars of the same kind of object...



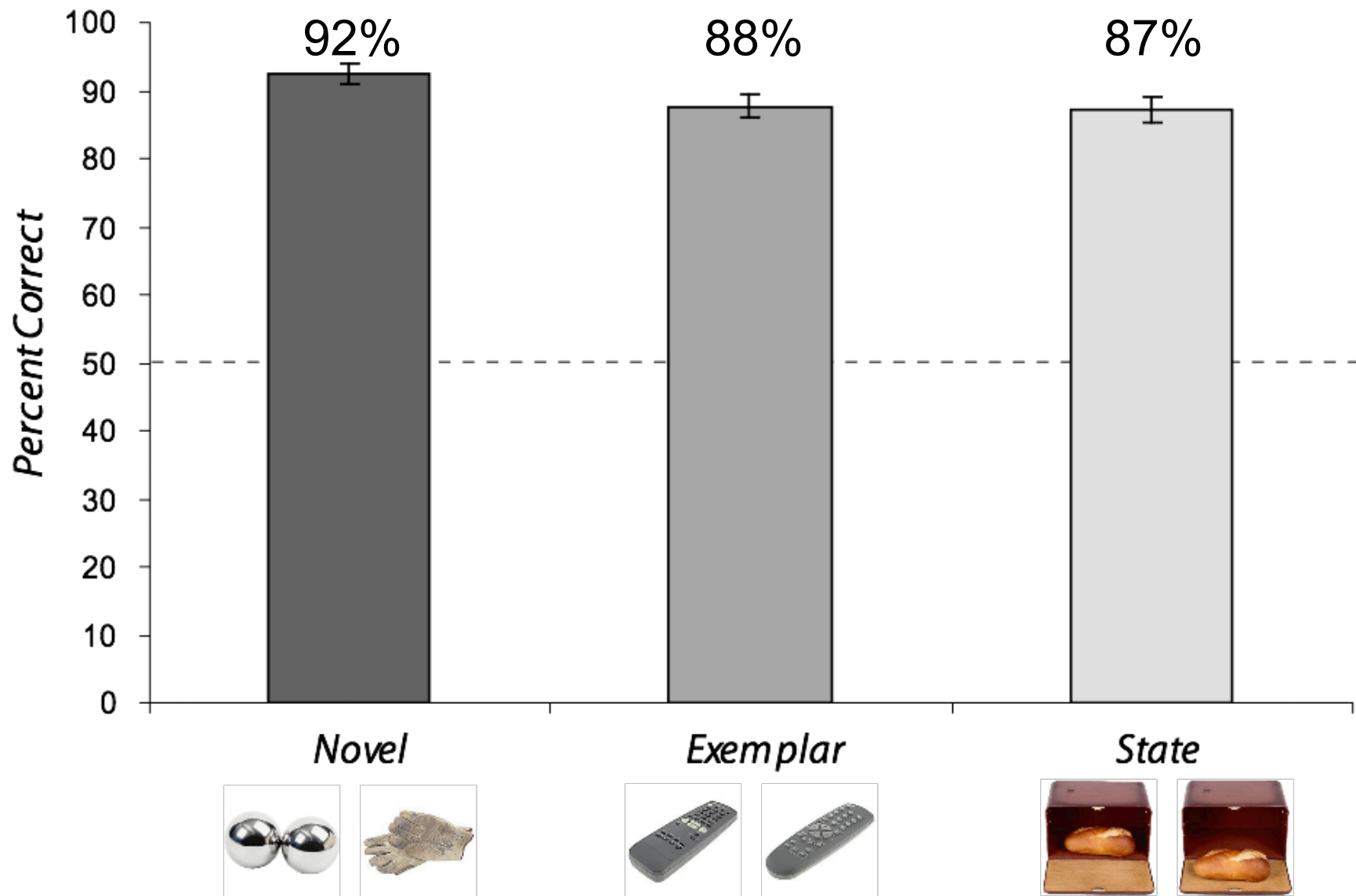
Different states of the same object...



Same object, different states



Recognition Memory Results

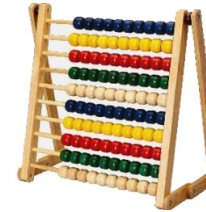


Brady, T.F., Konkle, T., Alvarez, G.A., & Oliva, A. (2008). Visual long-term memory has a massive storage capacity for object details. *Proceedings of the National Academy of Sciences, USA*, vol 105 (38), 14325-14329.

Examples of **Exemplar** Memory Tests



Examples of **State** memory test



Welcome to the

Visual Memory Game

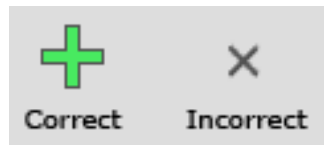
**A stream of images will be presented
on the screen for 1 second each.**

Your task:

Clap your hands (press a key) anytime you see an image you saw before.

Be attentive, repeats may be separated by many images !

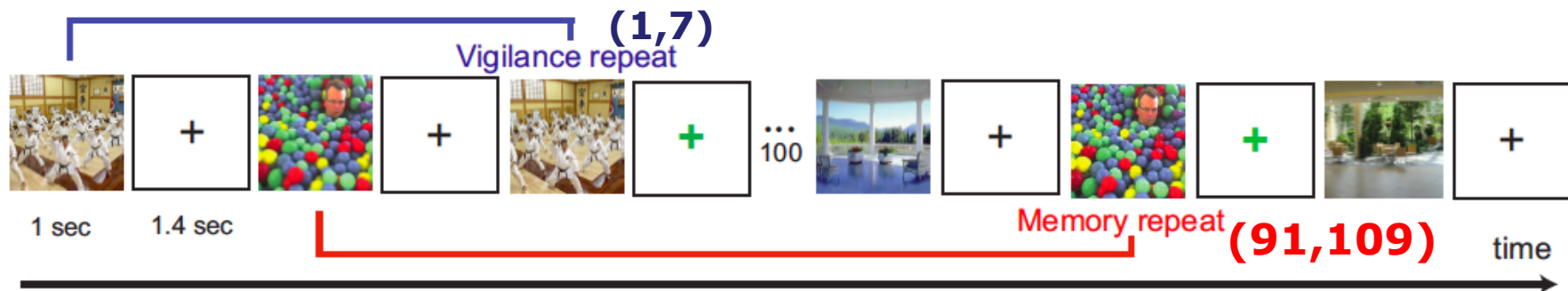
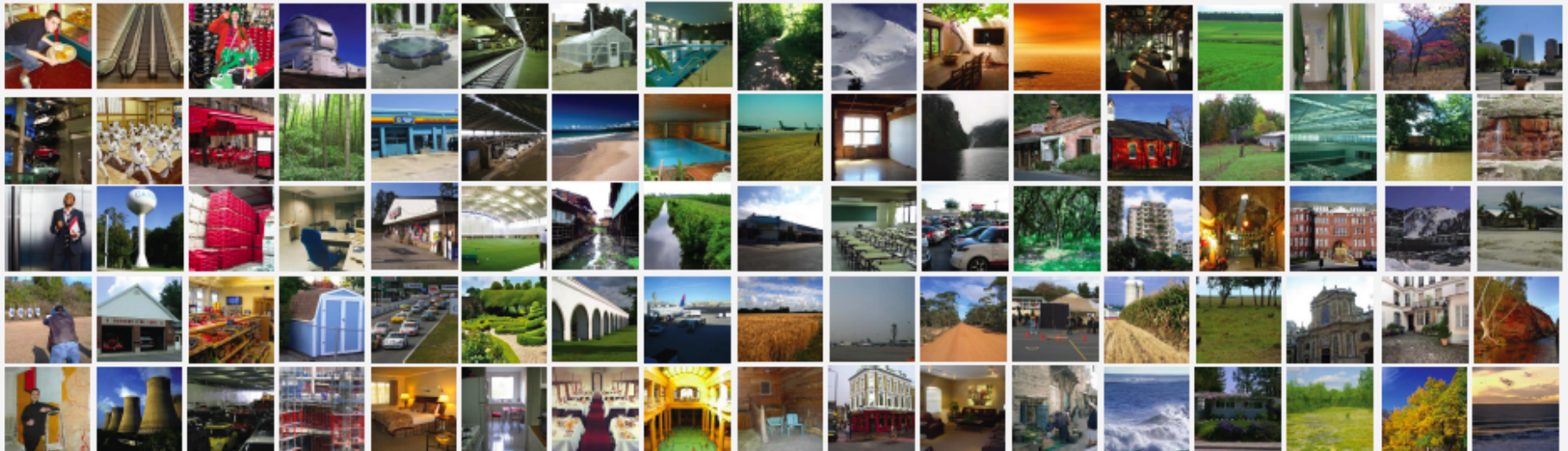
Whenever you press a key, you will get feedback:



You may exit the game at any time and you will be paid in proportion to your progress at that time

Start Game!

Visual Memory Game: Method

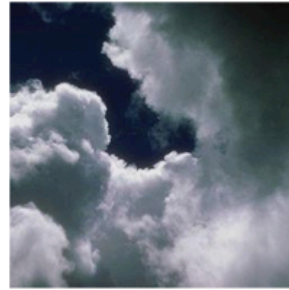


- Continuous repeat detection task
- **~ 10,000** unique images sampled from 900 scene categories (Standing, 1973; Brady et al., 2008)
- **2222** target images (memory repeats) whose repeats occurred **~ 91-109** after the first presentation

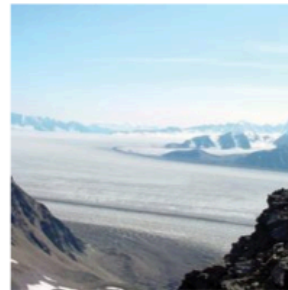
- Vigilance repeats every 1-7 images
- Each game level has 120 images
- N= 650 AMT workers
- **~ 80** scores per target images



Memorable
 Hit rate: 67/70
 False alarm rate: 4/80

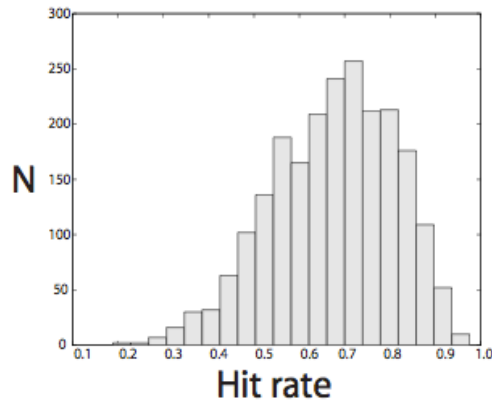


Average
 Hit rate: 59/81
 False alarm rate: 7/92



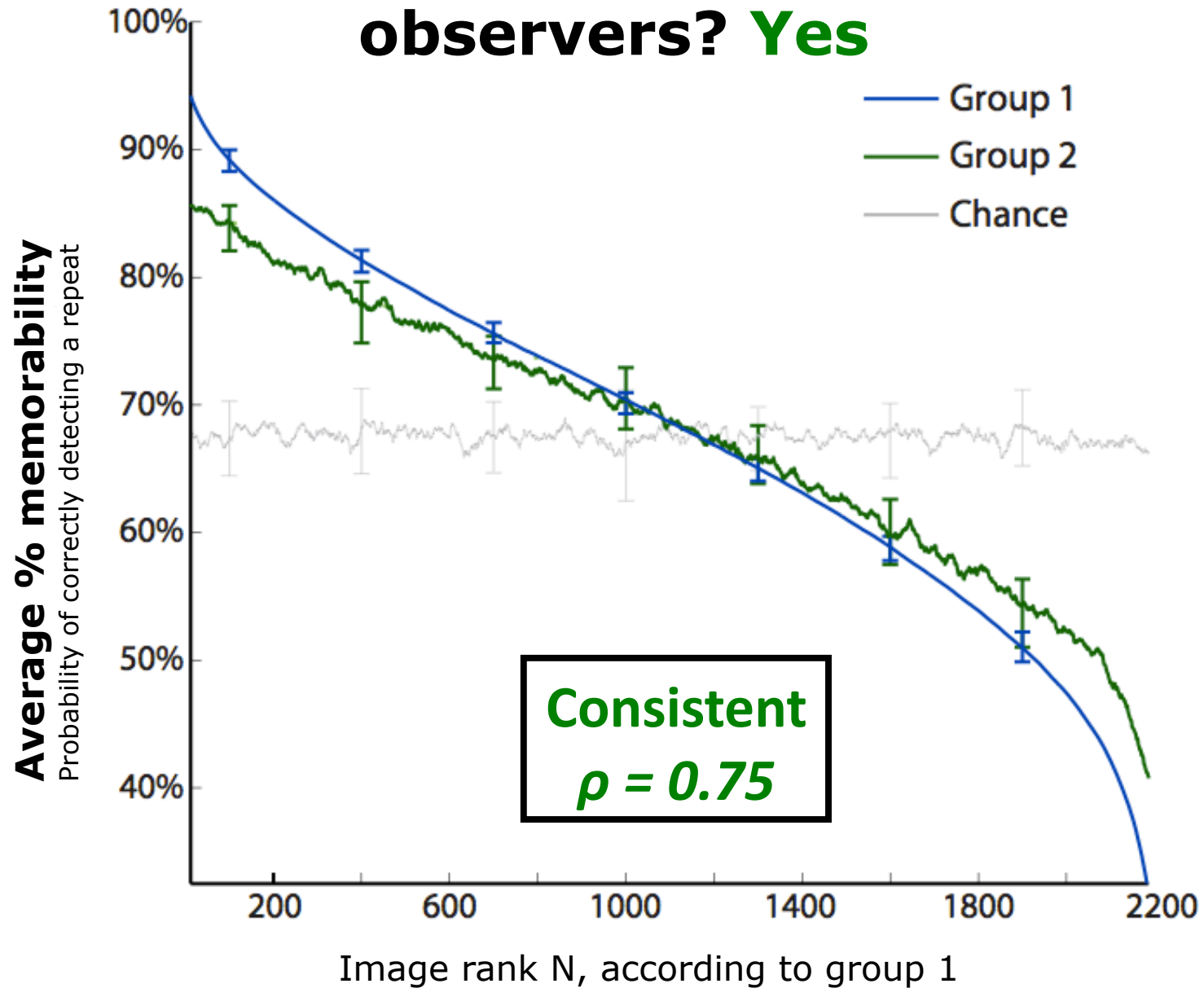
Forgettable
 Hit rate: 21/68
 False alarm rate: 3/82

Memorability
 Mean: 67.5%
 SD: 13.5%



**Large differences
 between images**

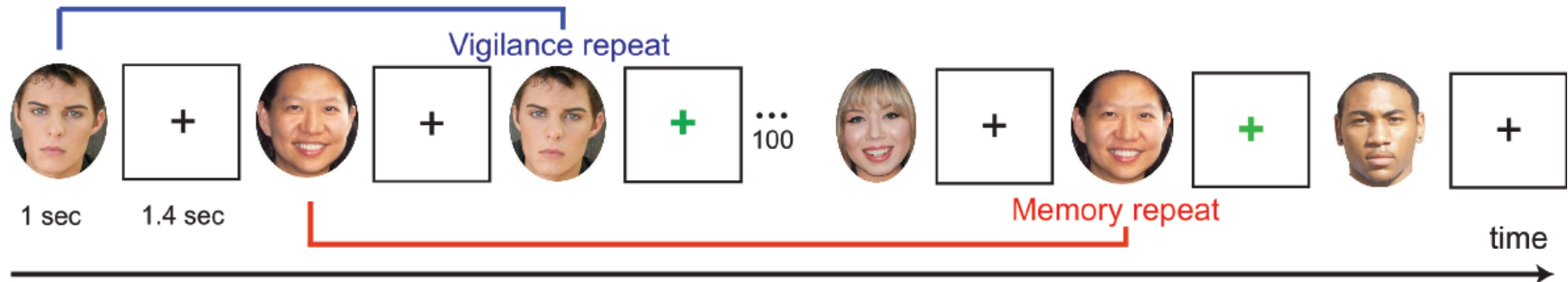
Is memorability consistent across different observers? **Yes**



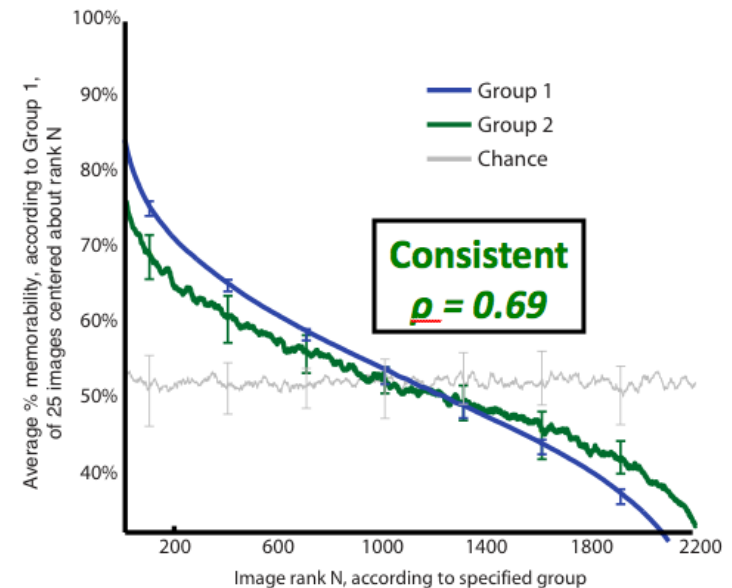
chance level calculated by randomly ordering the images on the x-axis

~ 80 scores per image

How consistent is memory within a single stimulus category?

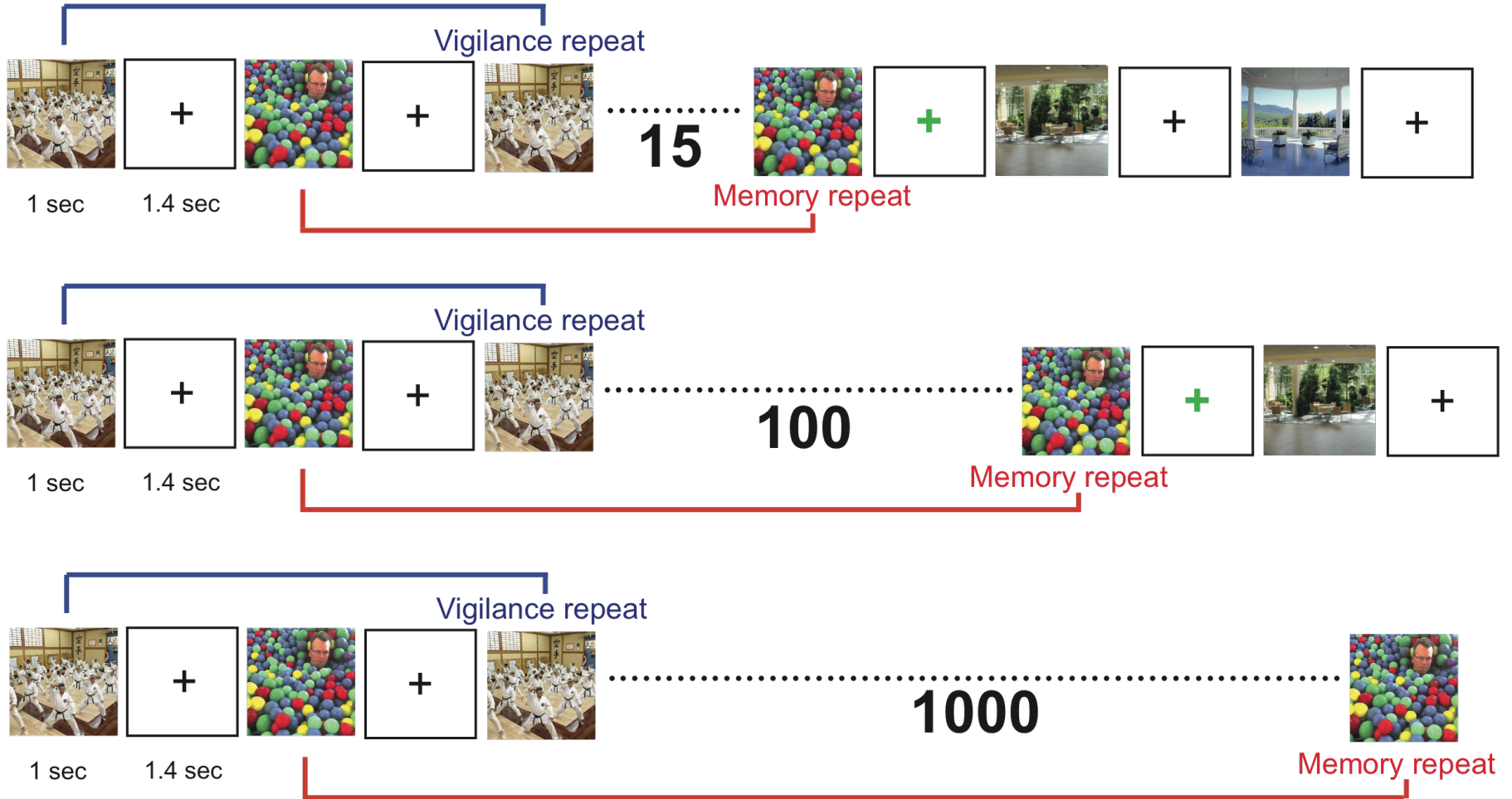


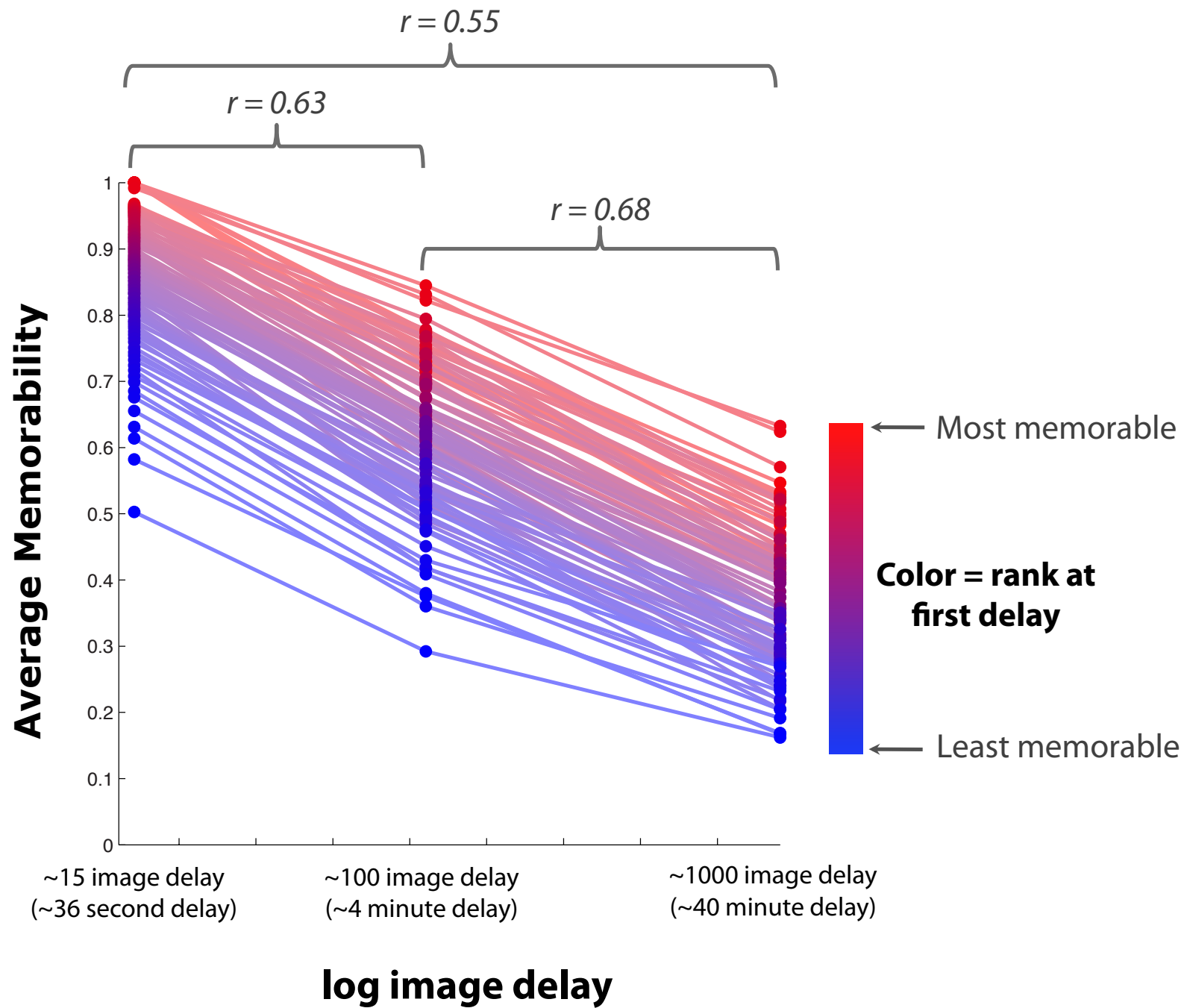
~10,000 unfamiliar faces, 2222 targets with ~ 80 memorability scores



Novel dataset: faces selection based on randomly generated first+last names following the distribution of the US census

Is memorability stable across time?



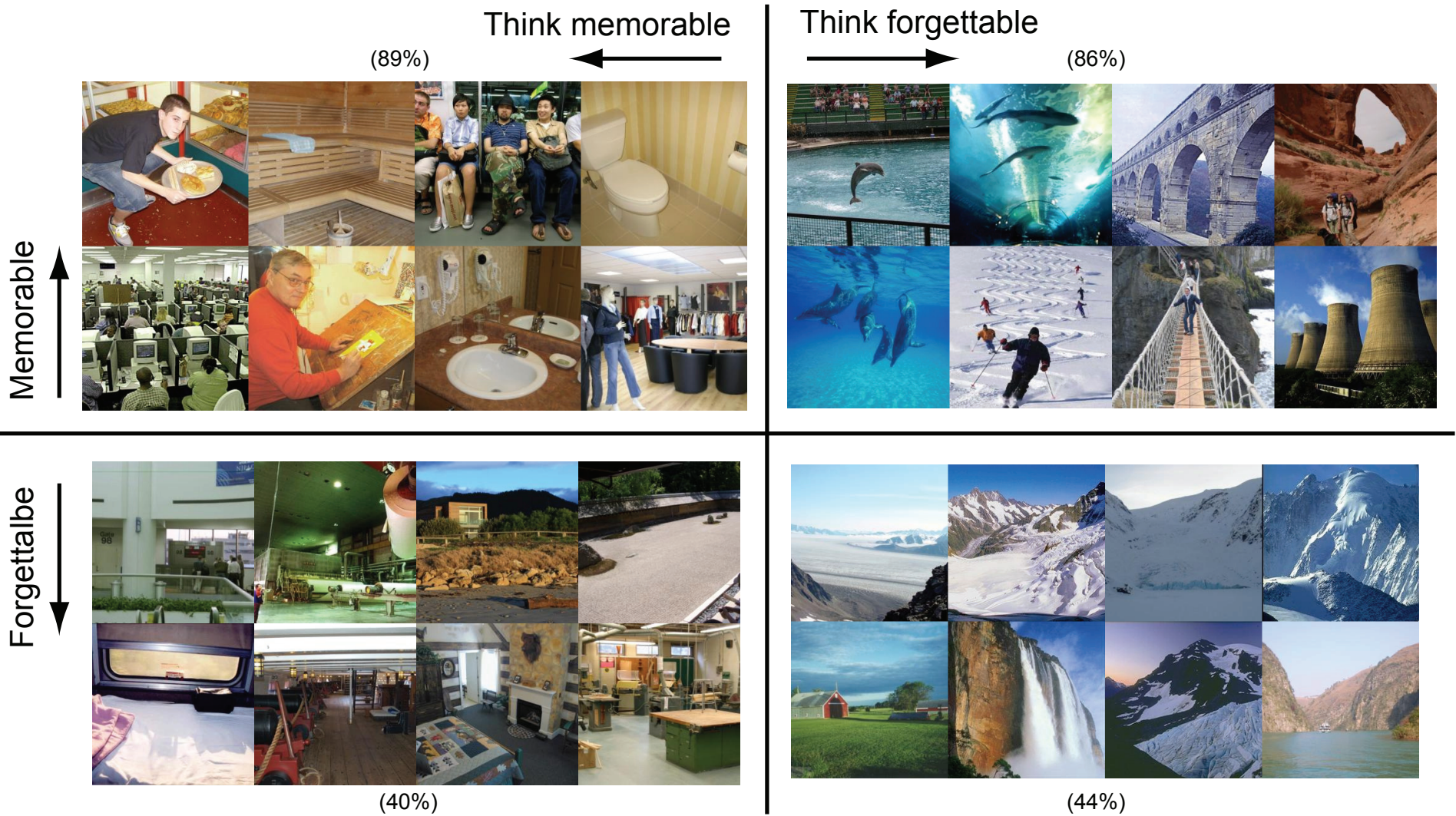


When do memorability differences arise?

At stage of encoding: Some images are encoded in less sufficient detail than others

Memorability rank changes very little over a wide portion of the memory trace!

Subjective judgments do not predict image memorability



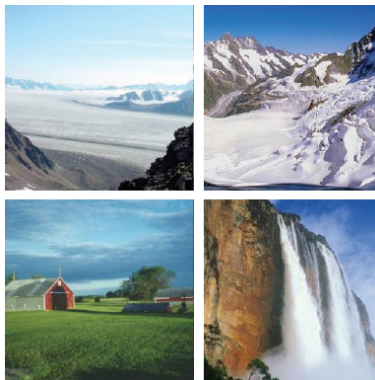
Can we estimate image memorability?

Humans

0.14 vs **0.89**

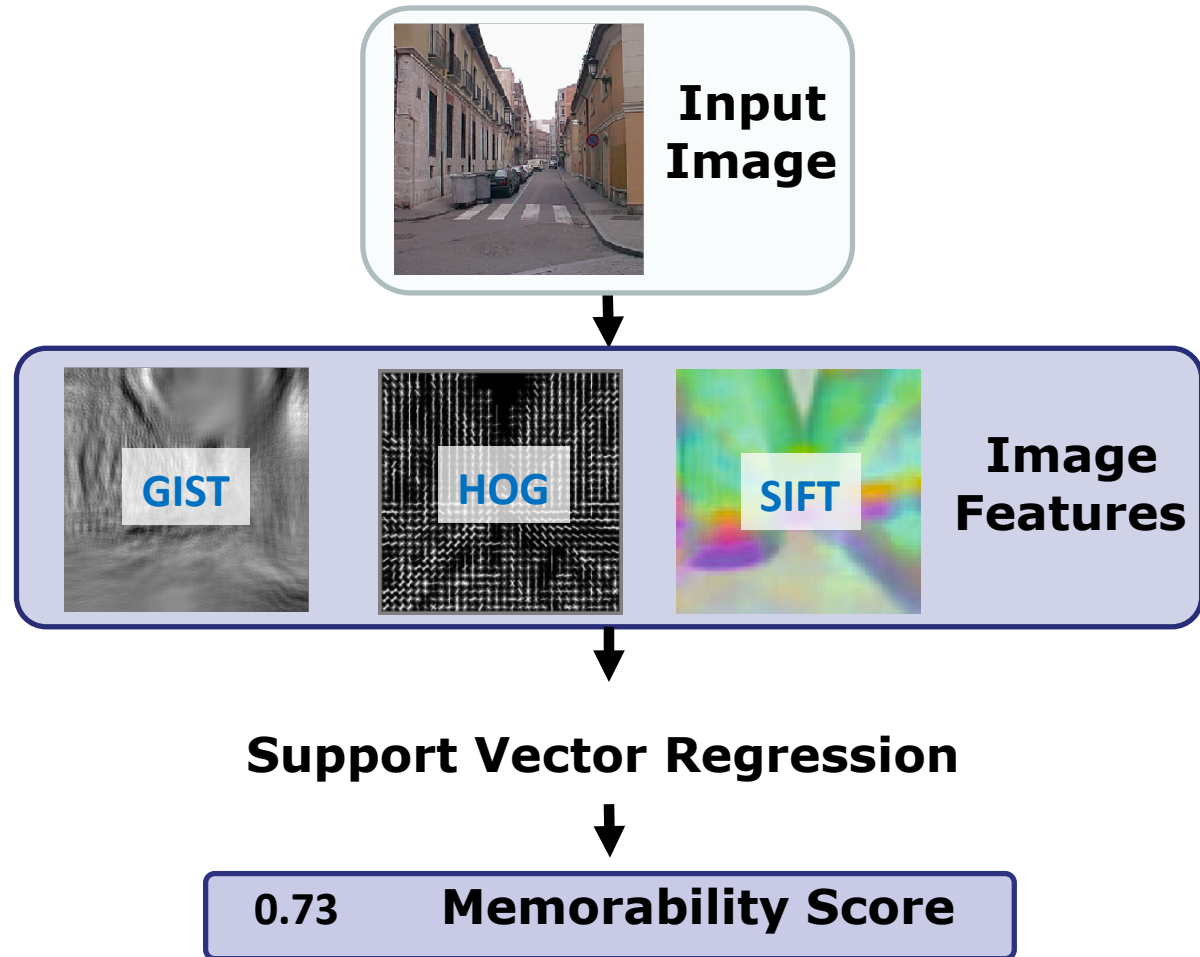


0.90 vs **0.44**



Human estimate
True Memorability

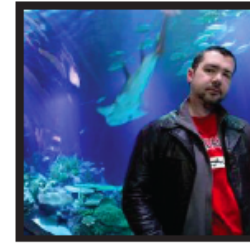
Computers



What makes an image memorable?

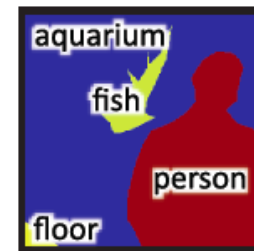
- **Intrinsic memorability?**

Are some images consistently more memorable than others, even across different observers and contexts?



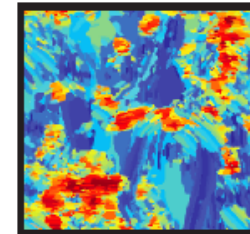
- **What image content matters?**

What image content (color, object, region) is driving memorability?

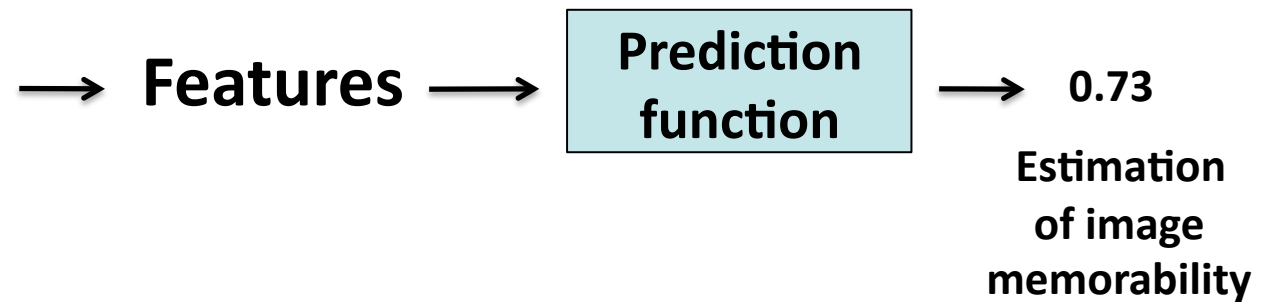


- **Can we predict it?**

Can we automatically predict an image's memorability from its image features?



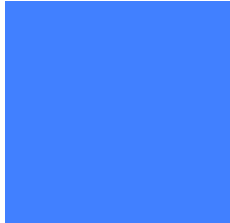
Prediction algorithm



The result of the regression will be a function that will take as input the features of an image and will output an estimate of the image memorability



Which features types predict memorability?



1) Simple scalar stats?

brightness, number of objects, mean hue

“Aquarium”

2) Scene category?

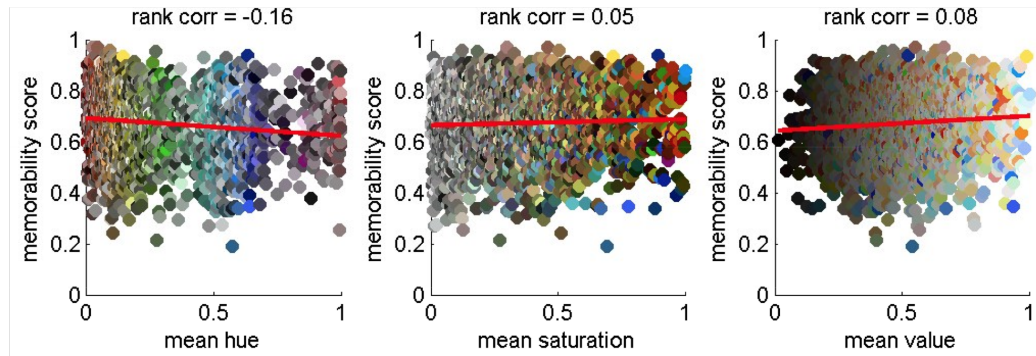
e.g. Aquarium, broadleaf forest, art studio



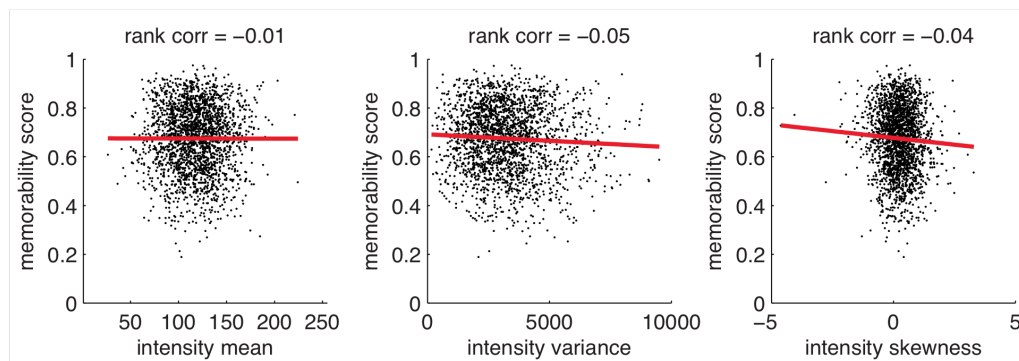
3) Object content?

number, size, and rough position of each object class

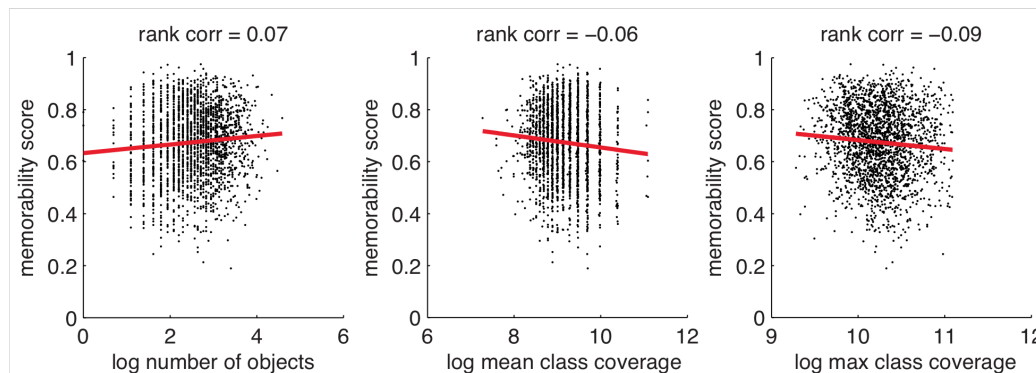
Simple, scalar summary statistics do not correlate well with memorability



← Color stats



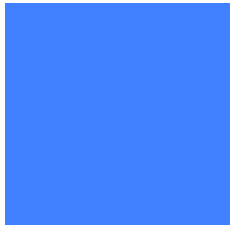
← Intensity stats



← Object stats



Which features types predict memorability?



1) Simple scalar stats?

brightness, number of objects, mean hue

$\rho < 0.16$

“Aquarium”

2) Scene category?

e.g. Aquarium, broadleaf forest, art studio



3) Object content?

number, size, and rough position of each object class

Scene features



“aquarium”



Which features types predict memorability?



1) Simple scalar stats?

brightness, number of objects, mean hue

$\rho < 0.16$

“Aquarium”

2) Scene category?

e.g. Aquarium, broadleaf forest, art studio

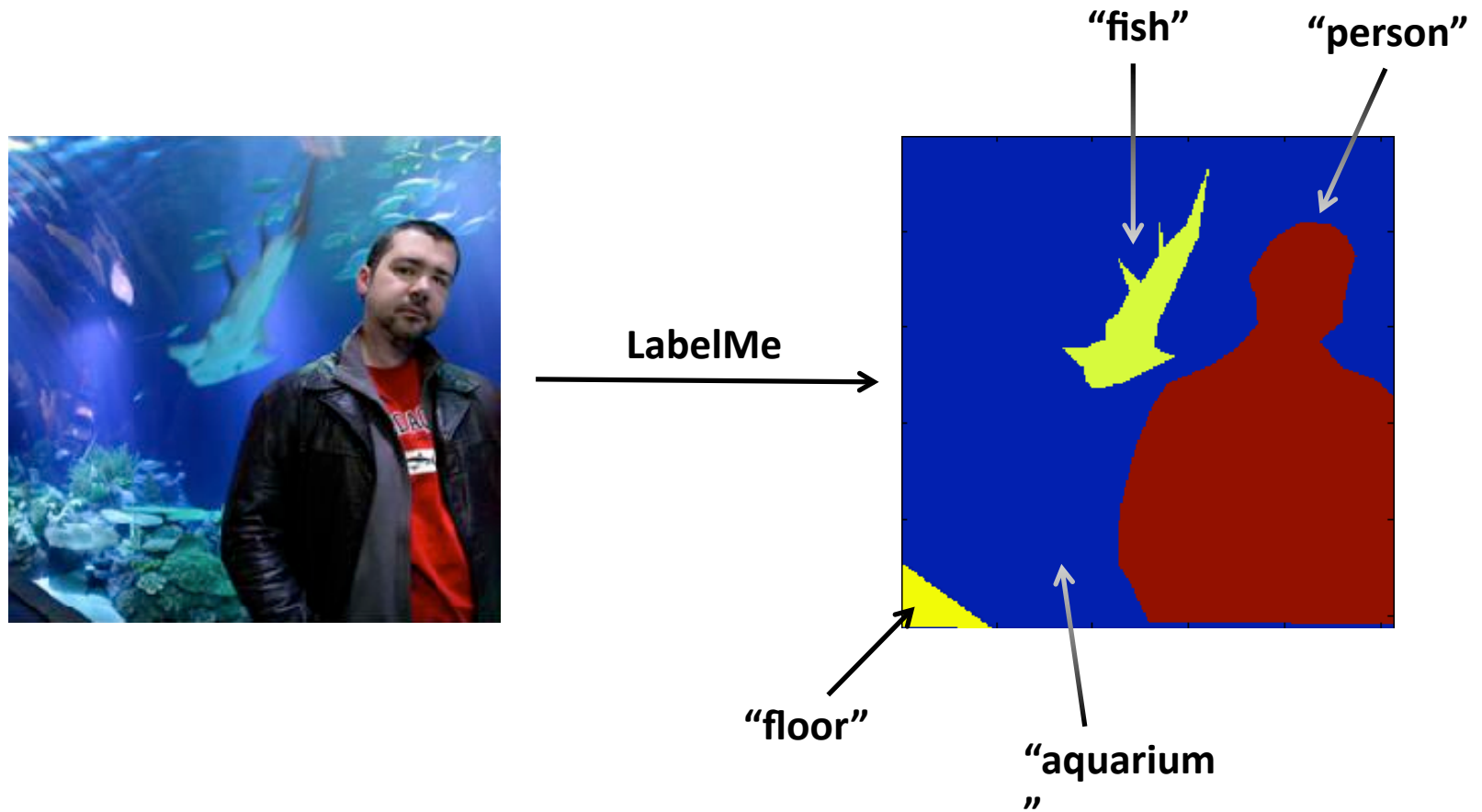
$\rho = 0.37$



3) Object content?

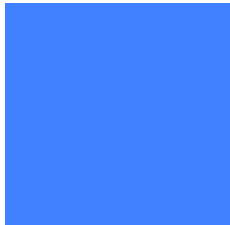
number, size, and rough position of each object class

Object features





Which features types predict memorability?



1) Simple scalar stats?

brightness, number of objects, mean hue

$\rho < 0.16$

“Aquarium”

2) Scene category?

e.g. Aquarium, broadleaf forest, art studio

$\rho = 0.37$



3) Object content?

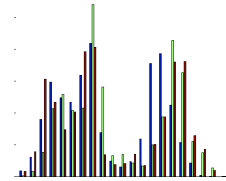
number, size, and rough position of each object class

$\rho = 0.48$

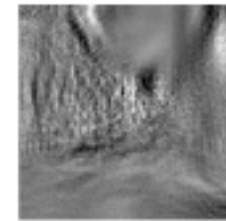
Global image features



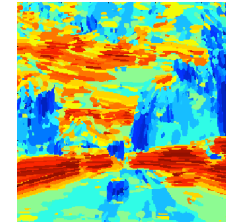
Pixel histograms



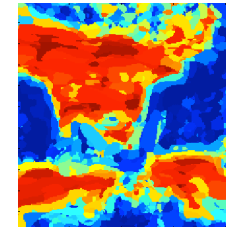
GIST
Gist



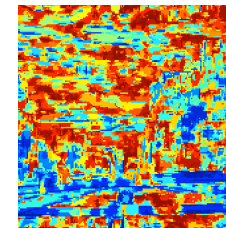
SIFT
Scale-invariant feature transform



HOG
Histogram of oriented gradients

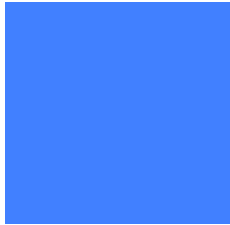


SSIM
Self-similarity





Which features types predict memorability?



1) Simple scalar stats?

brightness, number of objects, mean hue

$\rho < 0.16$

“Aquarium”

2) Scene category?

e.g. Aquarium, broadleaf forest, art studio

$\rho = 0.37$



3) Object content?

number, size, and rough position of each object class

$\rho = 0.48$

4) Global image features?

$\rho = 0.46$

Human consistency: $\rho = 0.75$

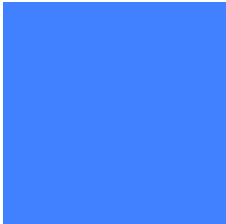


0) Human guessing?

asking people how memorable an image is

$$\rho = -0.02$$

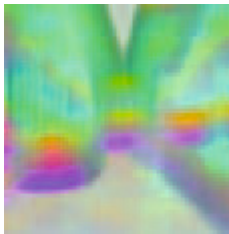
Can we estimate image memorability?



1) Simple scalar stats?

color, brightness, number of objects, mean hue

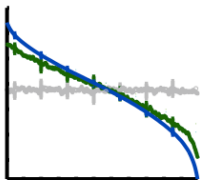
$$\rho < 0.16$$



2) Computer Model?

SIFT, HOG, GIST and SSIM

$$\rho = 0.46$$



3) Human objective estimation?

consistency across human subjects in memory game

$$\rho = 0.75$$

Human consistency

$\rho = 0.75$

Prediction by image features

$\rho = 0.46$



a) Most memorable images (86%)



a) Predicted most memorable (87%)



b) Typical images (74%)



b) Predicted typical memorability (68%)



c) Least memorable images (34%)



c) Predicted least memorable (52%)

Publications

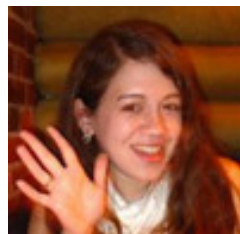
- Bainbridge, W., Isola, P., Blank, I., & Oliva, A. (2012). Establishing a Database for Studying Human Face Photograph Memory. Proceedings of the Cognitive Science Society.
- Isola, P., Xiao, J., Torralba, A., & Oliva, A. (2011). What makes an image memorable? Proceedings of the 24rd IEEE Conference on Computer Vision and Pattern Recognition (pp. 145-152).
- Isola, P., Parikh, D., Torralba, A., & Oliva, A. (2011). Understanding the Intrinsic Memorability of Images. Neural Information Processing Systems (NIPS).



Phillip Isola
Graduate student



Jianxiong Xiao
Graduate student



Wilma Bainbridge
Graduate student



Aditya Khosla
Graduate student

