

## Overview

### Background:

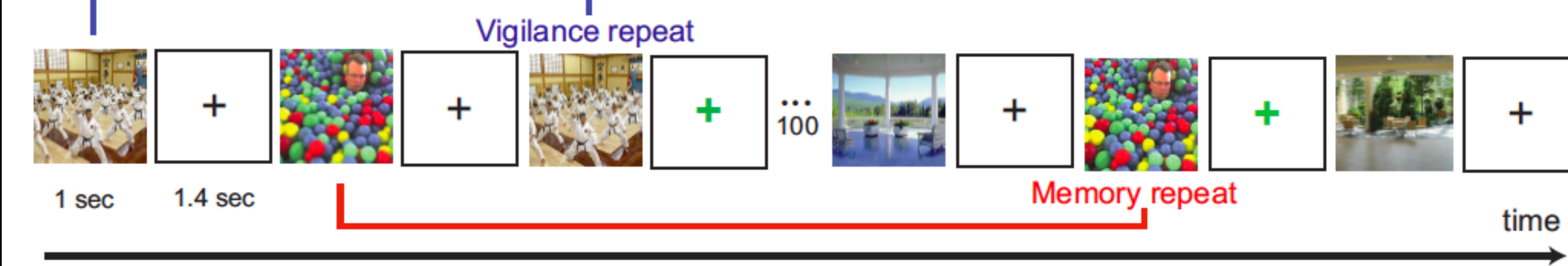


Image memorability is an *intrinsic* and *measurable* property! [Isola et al, CVPR 2011]

### Goal:

Find memorability of image regions automatically without manual annotation

### Method:

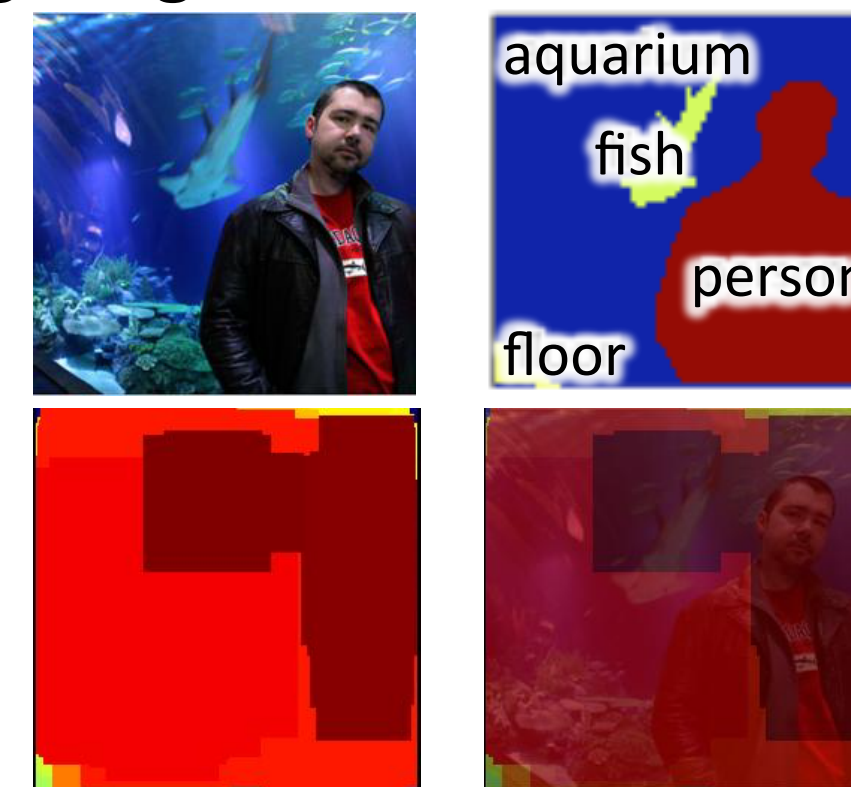
An *interpretable* model of memory as a noisy process composed of image regions

### Advantages over manual annotation:

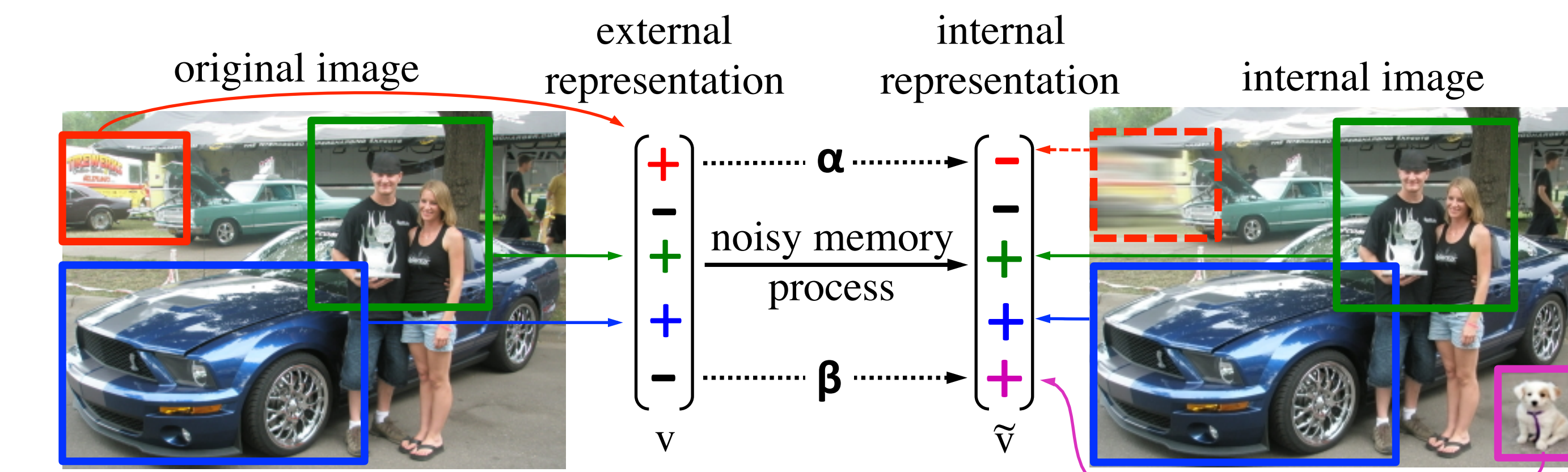
- Manual annotation of segments is expensive
- Granularity of annotations is not well defined

### Result:

- Same prediction performance as using ground truth image segments!
- Automatically generated memorability maps that correspond well with manually annotated images



## Algorithm



Expected distance of image  $j$ :

$$E(D_j|v_j) = \sum_{i=1}^N \alpha_i^{v_j(i)} * \beta_i^{1-v_j(i)} = v_j^T \vec{\alpha} + (-v_j)^T \vec{\beta} \propto_{rank} -s_j$$

Expected distance of all images:

$$E(D|v) = \begin{pmatrix} -v_1^T & \dots & -v_1^T \\ \vdots & & \vdots \\ -v_M^T & \dots & -v_M^T \end{pmatrix} \cdot \begin{pmatrix} \vec{\alpha} \\ \vec{\beta} \end{pmatrix} \propto_{rank} -\vec{s}$$

Ordinal Rank Regression with  $\alpha, \beta \in [0, 1]$ !

original image: image shown to observer

internal image: image retained in observers' brain

red: forgotten image region

blue, green: correctly retained image region in memory

pink: hallucinated image region

external representation ( $v$ ): observed image in terms of image region types,  $v \in \{0, 1\}^n$

internal representation ( $\tilde{v}$ ): image retrained in memory in terms of image region types,  $\tilde{v} \in \{0, 1\}^n$

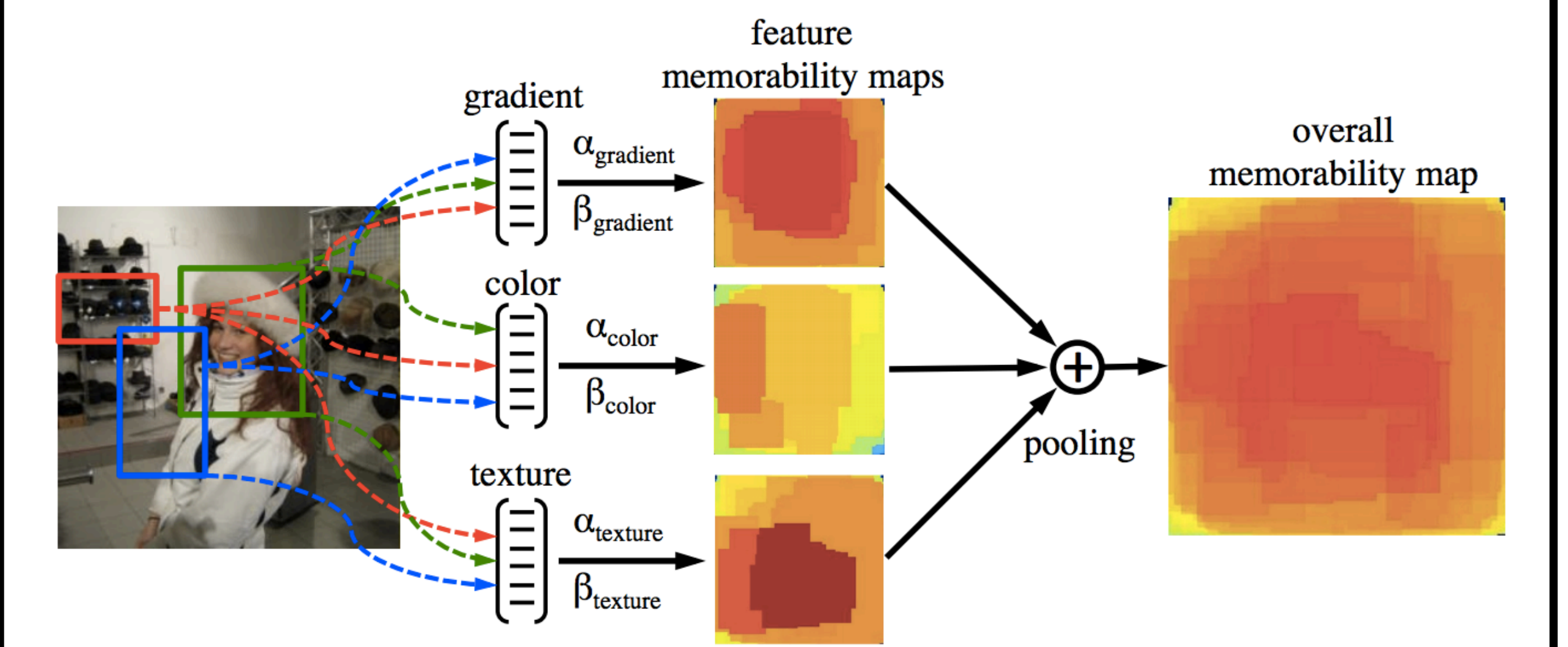
$\alpha_i$ : probability of forgetting image region of type  $i$

$\beta_i$ : probability of hallucinating image region of type  $i$

$s$ : image memorability score

$D$ : distance between internal and external representation i.e.  $\|v - \tilde{v}\|_1$

## Image Region Attributes



**Gradient**  
SIFT and HOG

**Color**  
Color Naming

**Texture**  
Local binary pattern

**Saliency**  
Eye-tracking based

**Shape**  
Image self-similarity

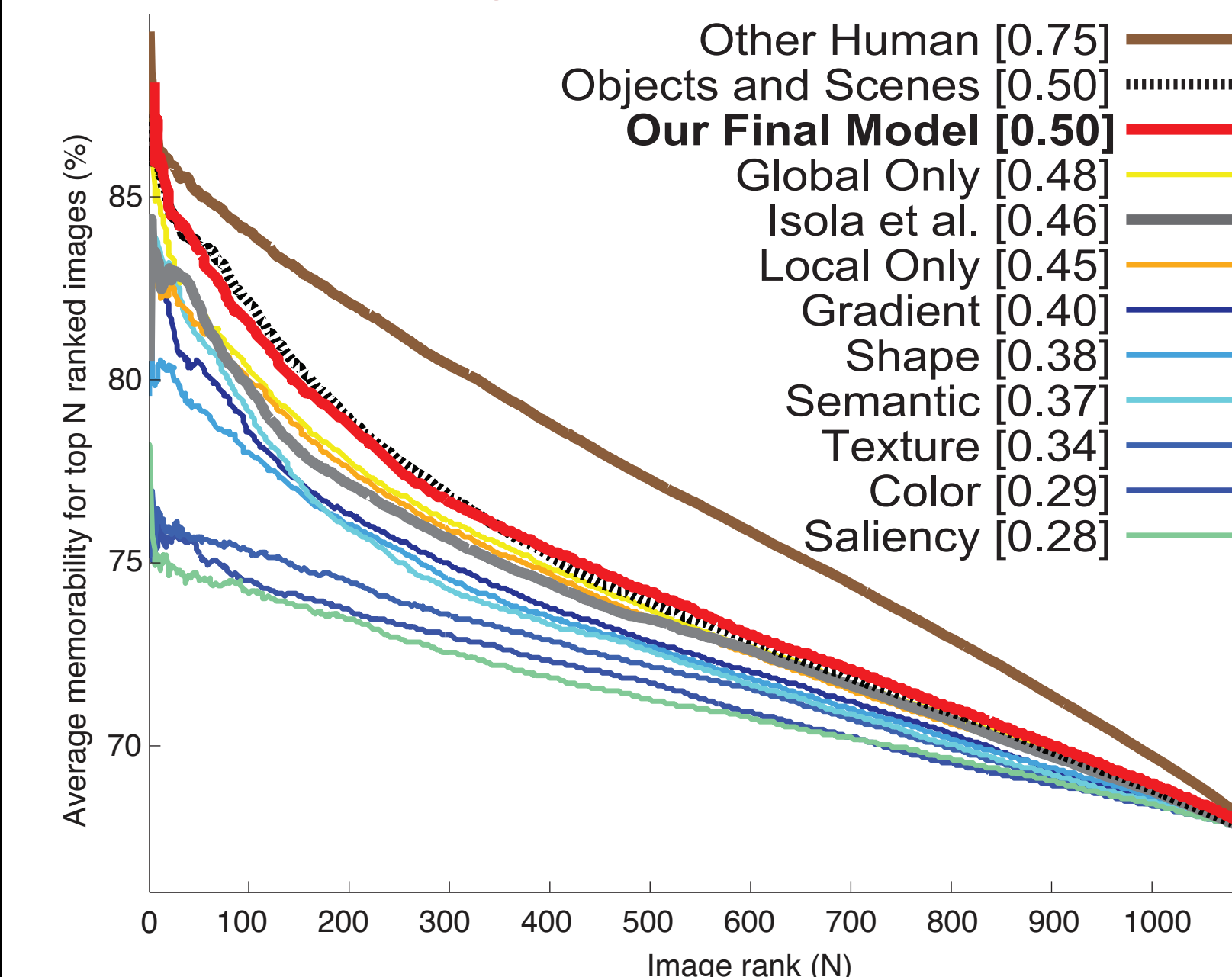
**Semantic**  
Object Bank

## Experiments

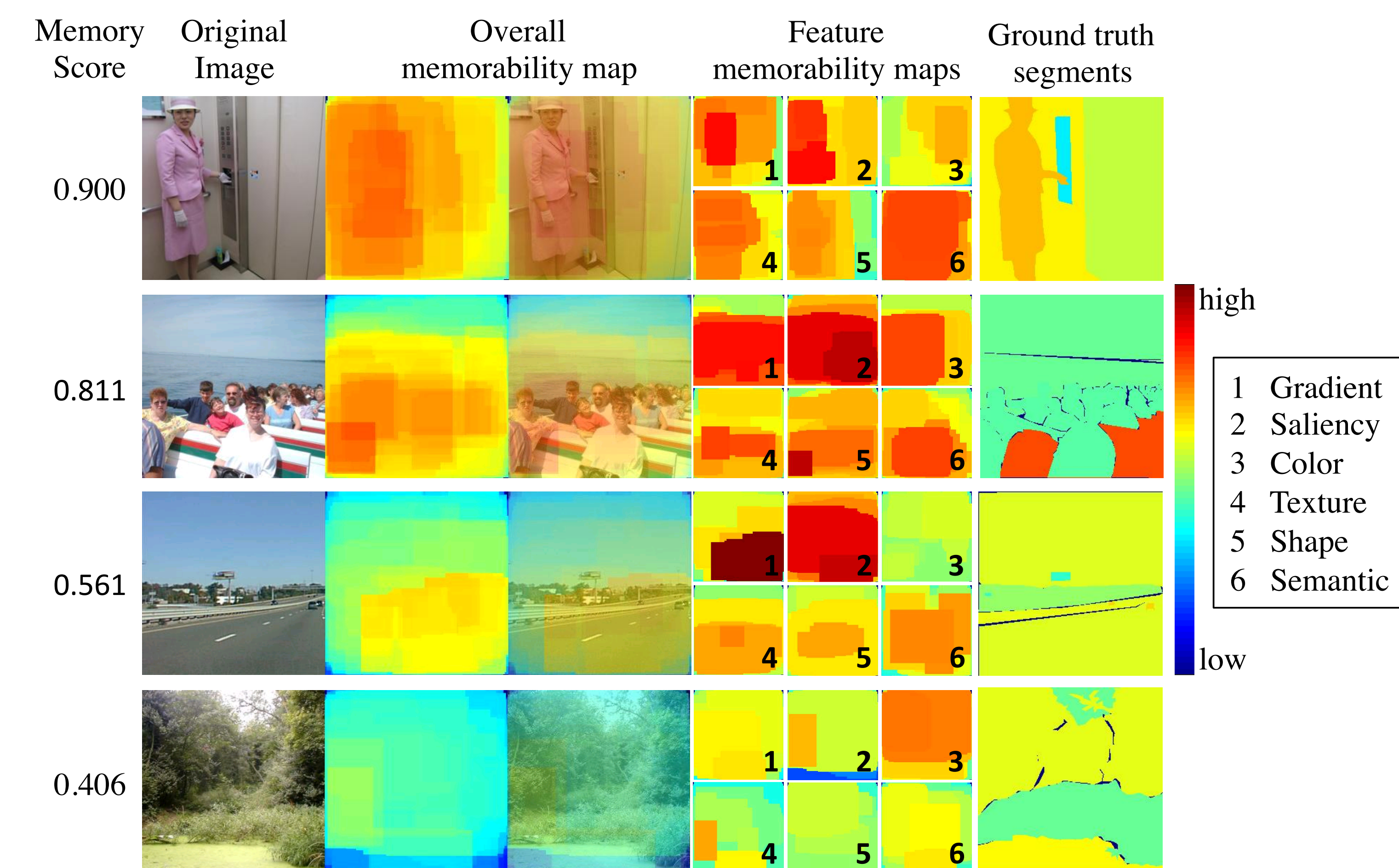
**Dataset:** Image memorability dataset with 2222 images, ~80 scores/image, 25 train/test splits (images sampled from SUN database [Xiao et al, CVPR 2010])

**Learning image region types:** k-means clustering with 256 dictionary size

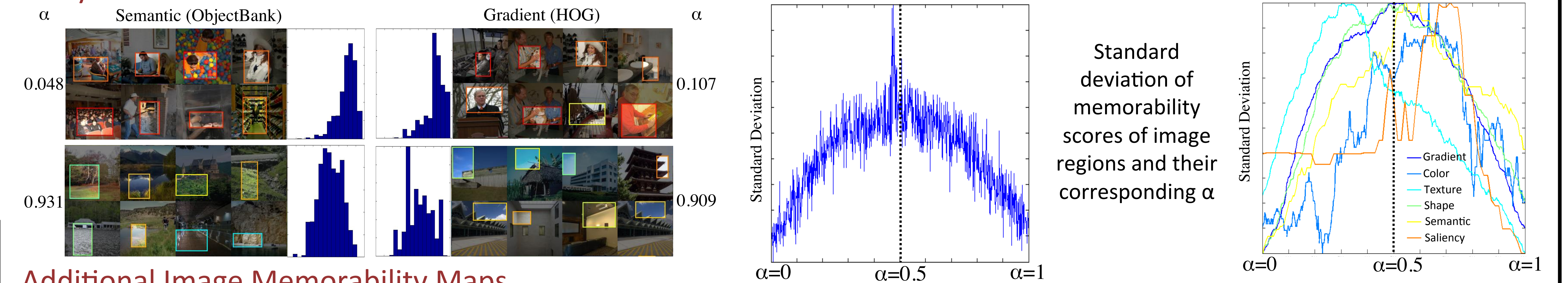
### Memorability Prediction



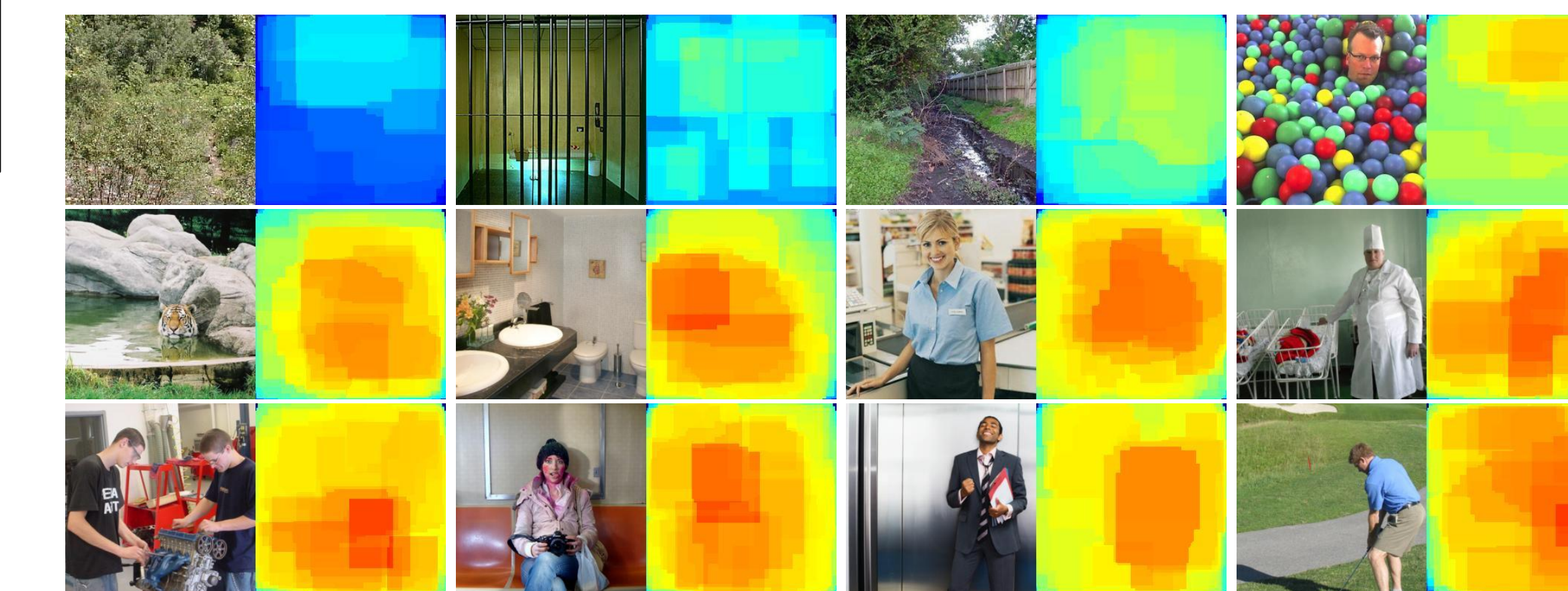
### Image Memorability Maps



### Analysis



### Additional Image Memorability Maps



## Conclusion

- We demonstrate an effective yet interpretable framework to automatically discover the memorability of image regions
- Future development of such automatic algorithms of image memorability could have many exciting and far-reaching applications in computer science, graphics, media, designs, gaming and entertainment industries in general