Image Compression Based on Visual Saliency at Individual Scales

Stella X. Yu¹ Dimitri A. Lisin²

¹Computer Science Department Boston College Chestnut Hill, MA

> ²VideoIQ, Inc. Bedford, MA

November 30, 2009

- 4 回 ト - 4 回 ト

Standard algorithms for lossy image compression:

- reduce entropy while minimizing difference from the original
- evaluate perceptual quality afterward
- Problem:
 - similarity to the original does not guarantee visual quality

- 4 回 ト - 4 三 ト

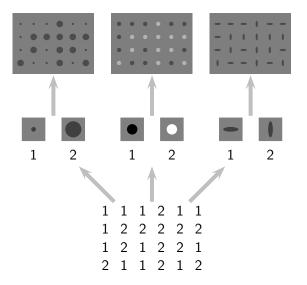
- ► Goal:
 - reduce entropy while preserving salient regions

Looking Without Seeing



▲圖 ▶ ▲ 臣 ▶ ▲ 臣

Experiments with Synthetic Images

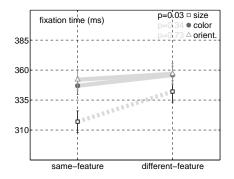


< ∃ >

-∢∃⇒

2

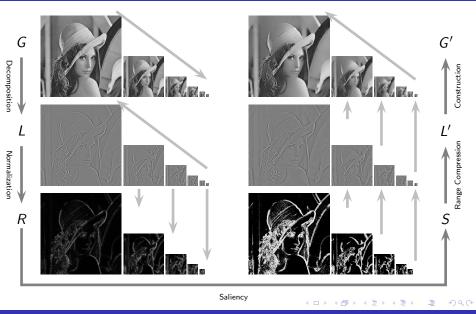
Results



(A) → (A) =

- Preview benefits for size
- No preview benefits for intensity or orientation
- Conclusion: visual attention is scale-selective

Multi-scale Saliency-Based Compression



Value-adaptive binning

- low values of the Laplacian are smoothed more than high values
- Scale-adaptive binning
 - range of the Laplacian naturally decreases with scale

< 同 > < 三 > < 三

finer scales are smoothed more than coarser scales

Test Images



Comparison with Signal-Based Compression



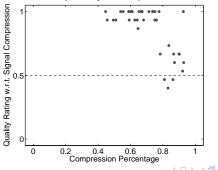
signal-based compression



our compression

Evaluation of Visual Quality

- ▶ 12 images, 3 compression levels, 15 (human) subjects
- Forced choice between a pair of images
- Brief exposure (1.2 s)
- Quality measured by percentage of votes



visual quality comparison:

Stella X. Yu and Dimitri A. Lisin. Image Compression Based on Visual Saliency at Individual Scales

Comparison with Wavelet Compression



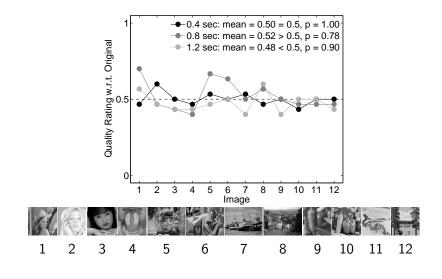
Daubechies wavelet compression



□ > < ∃ >

our compression

Comparison with the Originals



A ►

Compressed vs. Original



original image

our compression

<ロ> <回> <回> <回> <回> < 回>

Compressed vs. Original



original image

our compression

- 4 回 > - 4 回 > - 4 回

- Human vision study suggests scale-selectivity of attention
- Compression should preserve salient features at all scales
- Laplacian pyramid is used both as signal representation and as a saliency measure
- Range compression results in adaptive binning, reducing entropy while preserving visual quality
- Our algorithm can even enhance the visual quality of some images

- 4 同 ト 4 ヨ ト 4 ヨ ト

p specifies the fraction of pixels to be considered maximally salient scaling factor α controls saliency sharpness saliency map *S* at scale *s* is computed from *L* using a sigmoid with threshold *m*:

$$S_{s} = \left(1 + e^{-\frac{R_{s} - m_{s}}{\alpha}}\right)^{-1}, s = 1 : n$$

$$m_{s} = \arg\left\{\sum_{i} S_{s}(i; m_{s}) = p \cdot \sum_{i} 1\right\}, R_{s} = \frac{|L_{s}|}{\max(|L_{s}|)}$$
(2)

- - E > - E >

Given S, neighborhood radius r and range sensitivity factor β , generate a new Laplacian pyramid L' by spatially-variant range filtering of L:

$$L'_{s}(i) = \frac{\sum_{j \in N(i,r)} L_{s}(j) \cdot W_{s}(i,j)}{\sum_{j \in N(i,r)} W_{s}(i,j)}, L'_{n+1} = L_{n+1}, s = 1:n \quad (3)$$

$$W_{s}(i,j) = e^{-\frac{(L_{s}(j)-L_{s}(j))}{2\Theta_{s}(i)}},$$
(4)

$$\Theta_{s}(i) = (1 - S_{s}(i)) \cdot \left(\frac{\max(L_{s}) - \min(L_{s})}{\beta}\right)^{2}$$
(5)

- 4 回 ト - 4 回 ト