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Paper Sharing

Bottom-Up Saliency Detection Model Based on Human Visual Sensitivity and Amplitude Spectrum

Zhao Hongmiao

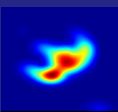
CVBIOUC

Ocean University of China

<http://vision.ouc.edu.cn/~zhenghaiyong>

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■ Author

- Fang Yuming
- 1984.10.
- lecturer
- Jiangxi University of Finance and Economics

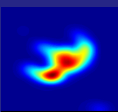


■ Citation

- IEEE TRANSACTIONS ON MULTIMEDIA
- Impact Factor: 1.776
- Citation Rate: 31

■ Highlights

- The human visual sensitivity and QFT amplitude spectrum
- Application: image retargeting



Saliency Detection Model

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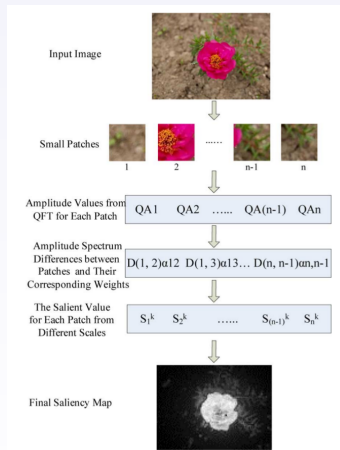
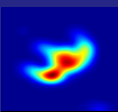


Figure : Framework of the proposed saliency detection model.



Saliency Value for Each Patch

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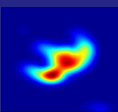
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- Two factors
 - The patch differences between this image patch and all other image patches in the input image $\implies \mathcal{D}_{(i,j)}$
 - The weighting for these patch differences $\implies \alpha_{ij}$

- Definition

$$\blacksquare S_i = \sum_{j \neq i} \alpha_{ij} \mathcal{D}_{(i,j)}$$



Differences Between Image Patches and Their Weighting to Saliency Value

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- Differences Between Image Patches
 - Quaternion Fourier Transform
 - $\mathcal{D}_{(i,j)} = \sqrt{\sum_m (\log(\mathcal{A}_m^i + 1) - \log(\mathcal{A}_m^j + 1))^2}$
- The weight for the patch difference between different patches¹

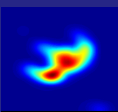
$$\mathcal{C}_s(f, e) = \frac{1}{\mathcal{C}_t(f, e)} \quad (1)$$

$$\mathcal{C}_t(f, e) = \mathcal{C}_0 \exp\left(\alpha f \frac{e + e_2}{e_2}\right) \quad (2)$$

$$e = \tan^{-1}\left(\frac{d}{v}\right) \quad (3)$$

$$\alpha_{ij} = \frac{1}{\mathcal{C}_0 \exp\left(\alpha f \frac{e + e_2}{e_2}\right)} \quad (4)$$

¹W.S. Geisler and J.S. Perry, “A real-time foveated multi-solution system for low-bandwidth video communication,” in Proc. SPIE, Jul. 1998, vol. 3299, pp. 294–305.



Patch Size and Scale for Final Saliency Value

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■ Patch Size

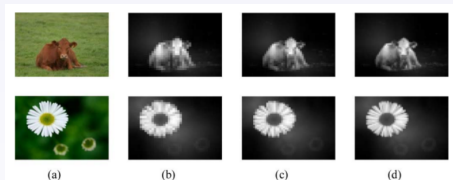
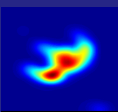


Figure : Original images and its different saliency maps with different patch sizes.

- The smaller image patch size, the more accurate saliency map we can get.
- Suitable patch size: consideration of fovea characteristics, saliency detection performance, and the computational complexity.



Patch Size and Scale for Final Saliency Value

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- Scale
 - the steerable pyramid algorithm
 - $S_i = \frac{1}{N} \sum_k S_i^k$

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- Database: MSRA²
- Comparison methods: PO, SR, MD, FG (temporary abbreviation)

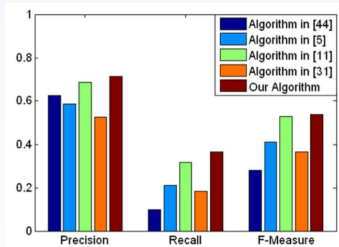


Figure : Experiment results for the comparison between the proposed model and others.

²T. Liu, J. Sun, N. Zheng, X. Tang, and H. Y. Shum, “Learning to detect a salient object,” in Proc. IEEE Int. Conf. Computer Vision and Pattern Recognition, 2007.



Conclusions

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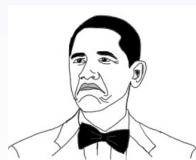
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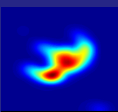
■ Advantages

- Image Patch \implies Superpixels
- $\mathcal{D}_{(i,j)}$ is novel
- Steerable pyramid algorithm

■ Disadvantages

- Accuracy
- Computational complexity





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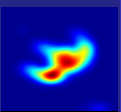
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■ Conclusion in frequency domain

Model	Method	Model	Method
SR	Residual Spectrum	HFT	SSS
PFT	Only Phase Spectrum	SDSP	log-Gabor
SIG	DCT+Sign	FT	DoG



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