



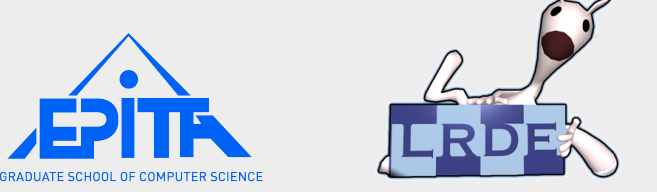
Salient Level Lines Selection Using the Mumford-Shah Functional

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Main Contributions

An efficient image simplification method:

- minimizing the Mumford-Shah functional on the tree of shapes [1];
- a fast greedy algorithm providing a local optimal solution.

A hierarchical image simplification:

- a variant of morphological shaping [2];
- a saliency map representing a hierarchical image simplification.

Effective results: color images pre-segmentation



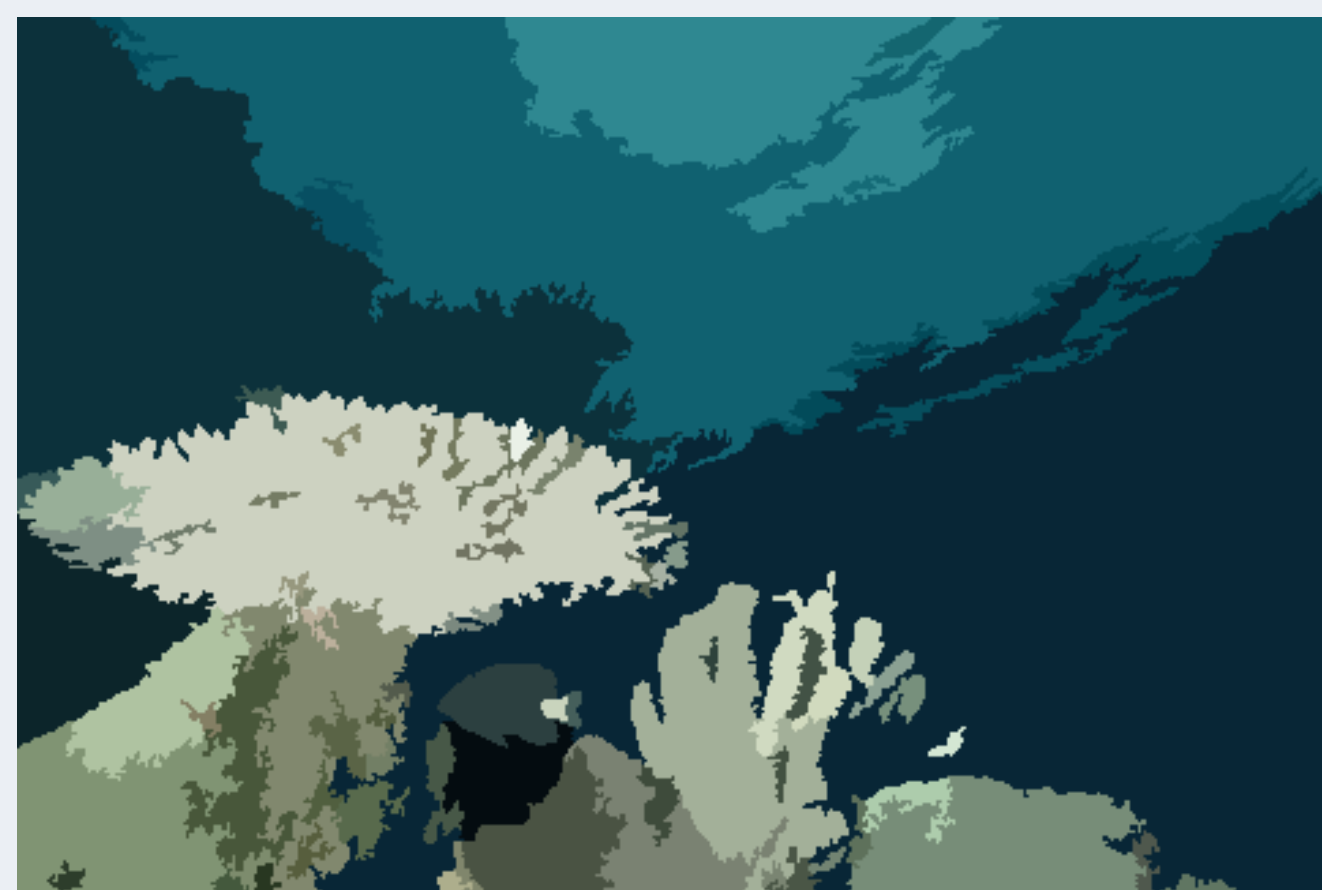
Input color image.



Input color image.



133 regions.



84 regions.

Effective results: hierarchical image simplification



Original image.



Saliency map.



Less simplified.



More simplified.

Basic idea

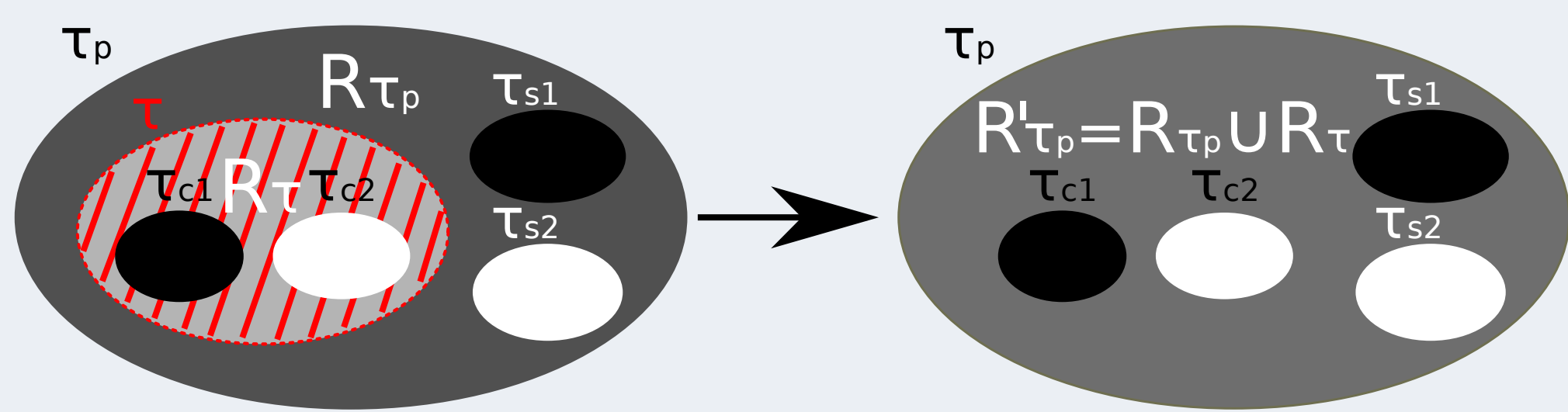
Minimize the piecewise-constant Mumford-Shah functional [3]...
...subordinated to the tree of shapes [1, 4] of an image f .

Piecewise-constant Mumford-Shah functional E :

$$E(f, \partial R) = \iint_R (\tilde{f}|_R - f)^2 dx dy + \nu |\partial R|$$

Tree of shapes \mathcal{T} : inclusion tree of level lines

Basic operation: level lines removal \Rightarrow region merging



Removing $\tau \Rightarrow R'_{\tau_p} = R_{\tau_p} \cup R_{\tau}$, iff

$$\Delta E_{\tau} = \frac{S^2(f, R_{\tau})}{|R_{\tau}|} + \frac{S^2(f, R_{\tau_p})}{|R_{\tau_p}|} - \frac{S^2(f, R'_{\tau_p})}{|R'_{\tau_p}|} - \nu |\partial \tau| < 0,$$

where $S(f, R) = \int_R f dx dy$.

Fast algorithm

Removing a node τ impacts its relatives \Rightarrow the removal order is critical.

1) Algorithm of Ballester et al. [5]:

remove the node τ that decreases the most E
 \rightarrow prohibitive $O(N^2)$ time complexity.

2) Our proposed algorithm:

the removal order is based on a meaningfulness criterion...
... used meaningfulness: average of gradient magnitude along $\partial \tau$.

Input: tree \mathcal{T} of image f ; Output: simplified tree \mathcal{T}'

- Initialization: set $\mathcal{T}' = \mathcal{T}$.
- Step 1: sort $\{\tau | \tau \in \mathcal{T}\}$ in increasing order \mathcal{O} of shape meaningfulness.
- Step 2: propagate $\{\tau\}$ following the order \mathcal{O} , remove τ , if $\Delta E_{\tau} < 0$.

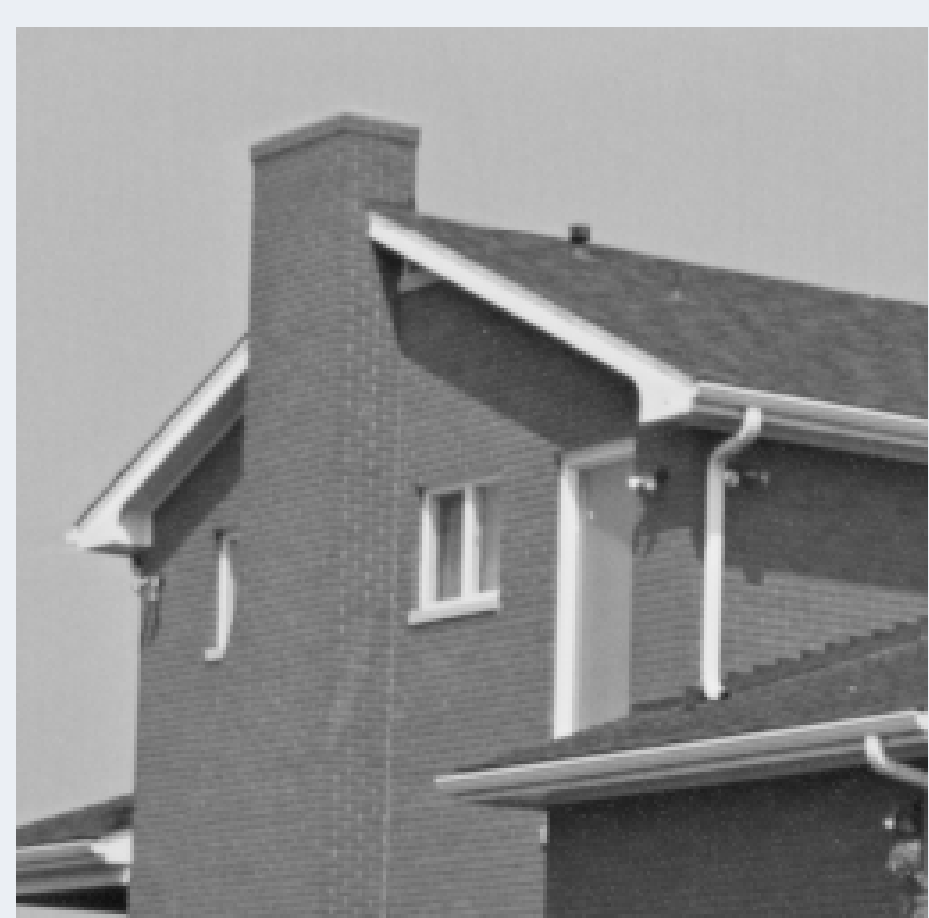
\mathcal{T}' : a locally optimal solution of $\min_{\mathcal{T}'} E(f, \partial \mathcal{T}')$.

2 or 3 iterations of step 2 might be desired \Rightarrow quasi-linear complexity

3) A variant of the proposed algorithm: compute $\nu_{\min}(\tau)$ instead of a fixed ν

- ν_{\min} : a non-increasing attribute,
- Attribute filtering \Rightarrow morphological shaping [2],
- Saliency map [2].

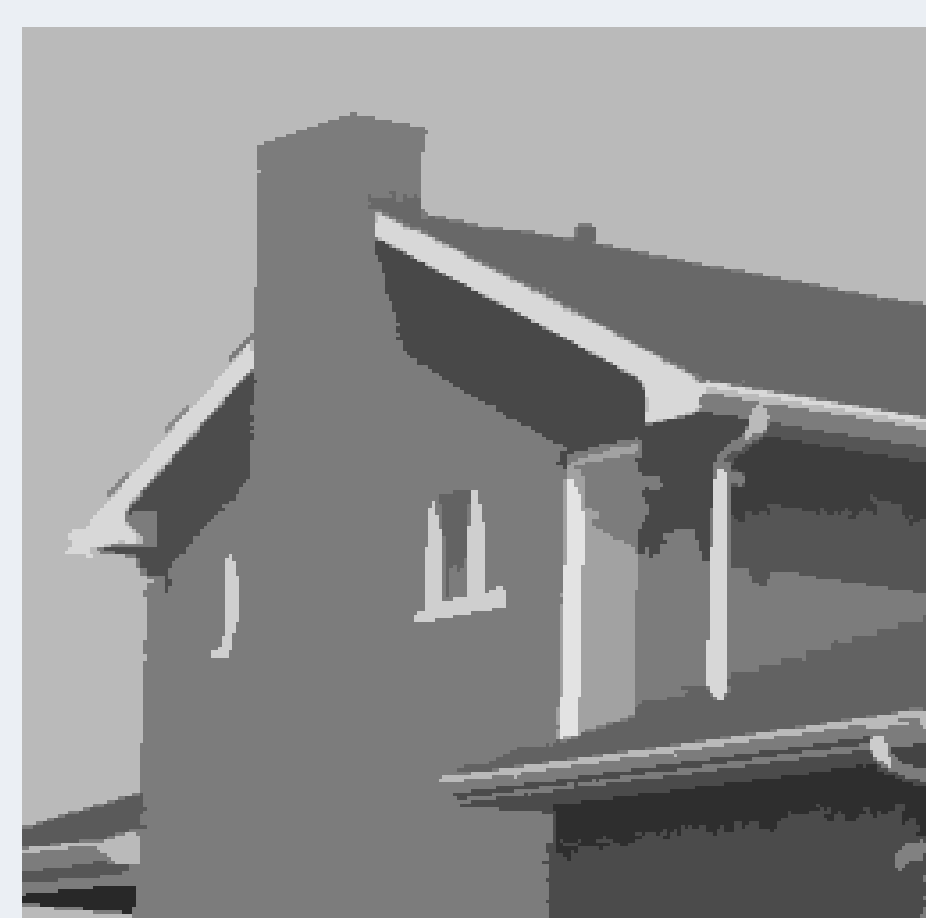
Comparison with the approach of Ballester et al. [5]



(a) Input image.

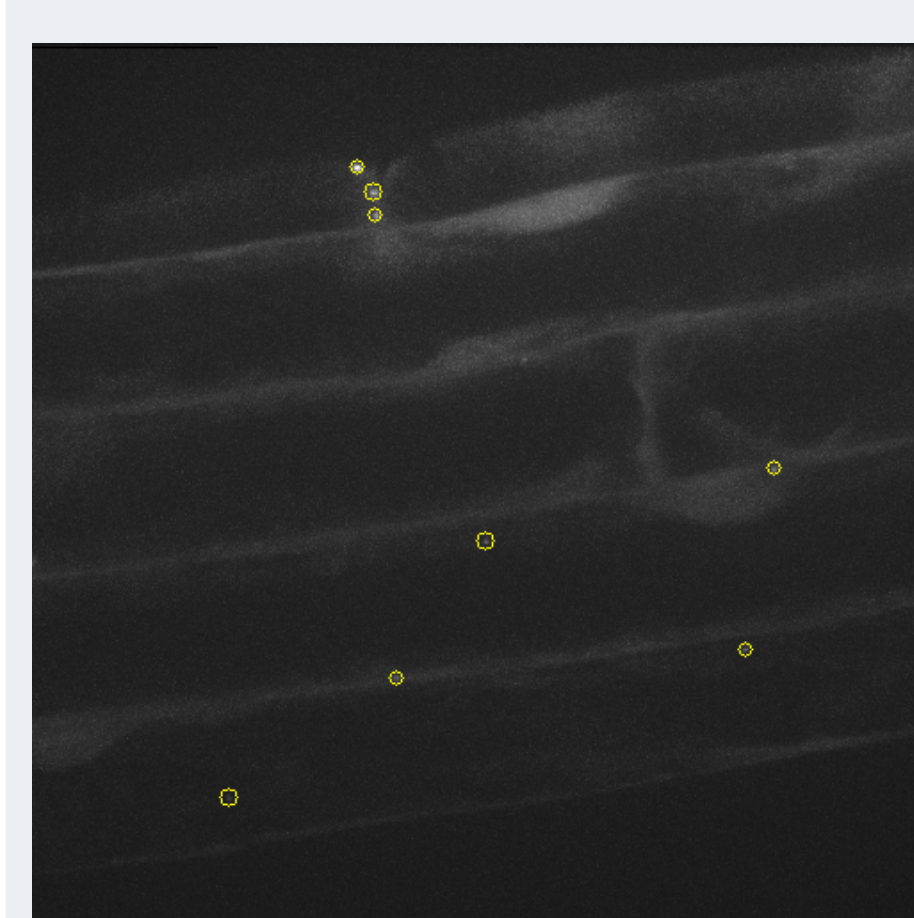


(b) Ballester, $\nu = 1k$, ($t = 4s$).

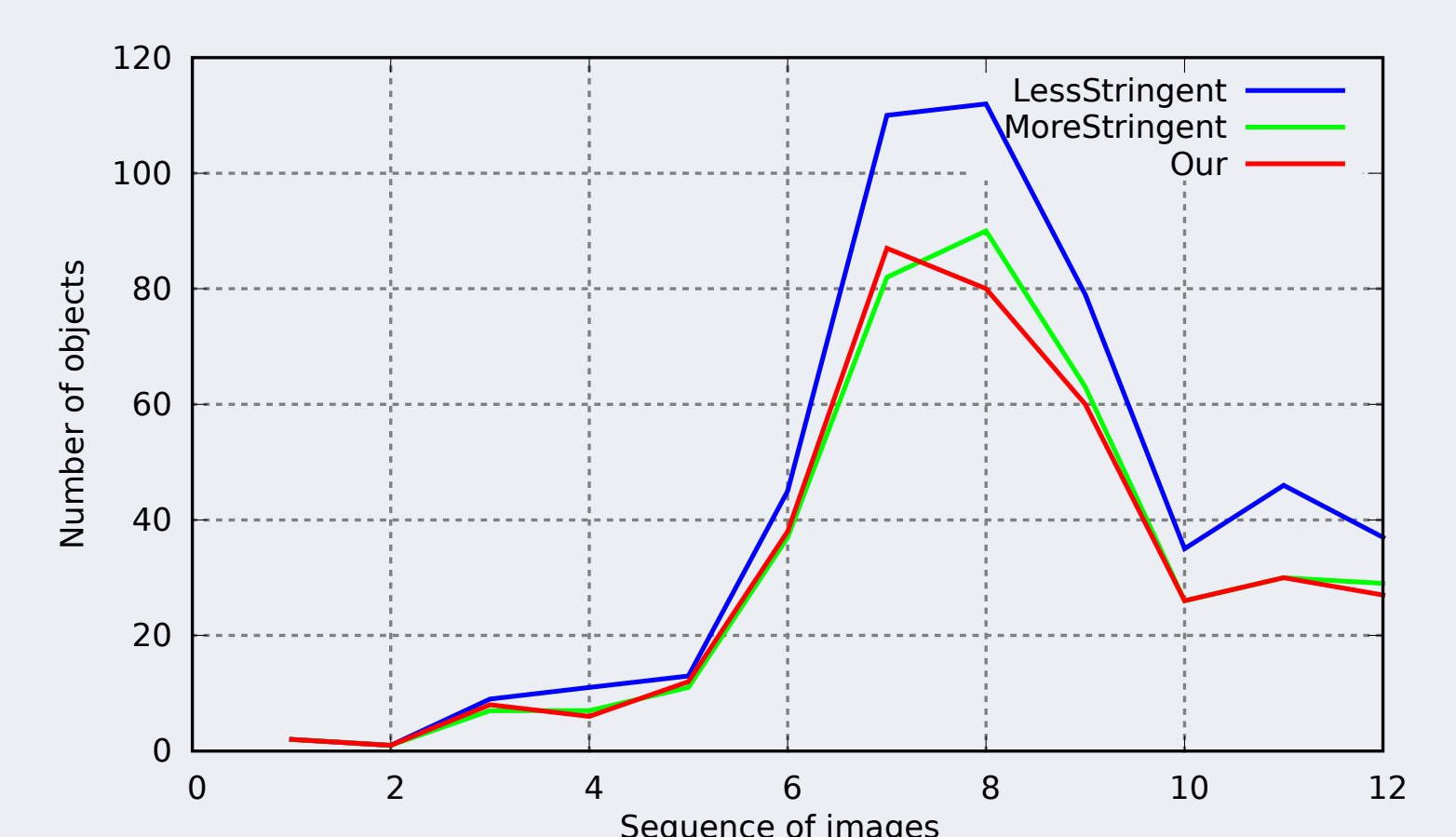


(a): $E = 1.24 \times 10^8$; (b): $E = 1.28 \times 10^7$, PSNR = 27.3; (c): $E = 1.15 \times 10^7$, PSNR = 28.7.

One application to autophagosome counting



Autophagosome counting.



Benchmark on a sequence of images.

Autophagosome in cellular images: small, round, bright objects.

References

- [1] P. Monasse and F. Guichard, "Fast computation of a contrast-invariant image representation," *IEEE Trans. on Image Processing*, vol. 9, no. 5, pp. 860–872, 2000.
- [2] Y. Xu, T. Géraud, and L. Najman, "Morphological Filtering in Shape Spaces: Applications Using Tree-Based Image Representations," in *ICPR*, pp. 485–488, 2012.
- [3] D. Mumford and J. Shah, "Optimal approximations by piecewise smooth functions and associated variational problems," *Communications on Pure and Applied Mathematics*, vol. 42, no. 5, pp. 577–685, 1989.
- [4] T. Géraud, E. Carlinet, S. Crozet, and L. Najman, "A quasi-linear algorithm to compute the tree of shapes of n d images," in *ISMM*, pp. 98–110, 2013.
- [5] C. Ballester, V. Caselles, L. Igual, and L. Garrido, "Level lines selection with variational models for segmentation and encoding," *JMIV*, vol. 27, pp. 5–27, 2007.