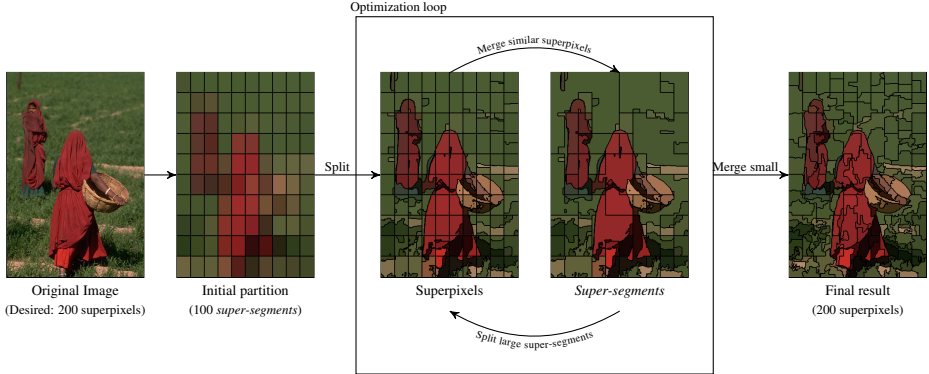


# SMURFS: Superpixels from Multi-scale Refinement of Super-regions

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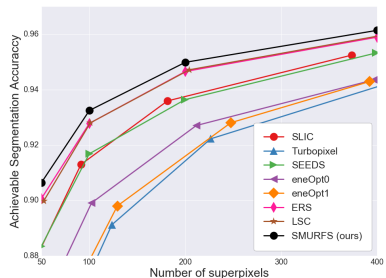
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**Figure 1:** Overview of our algorithm. Iterative refinement over two scales of regions yields increasingly more robust superpixels that better capture global image features.

Here we present a new superpixel algorithm: Superpixels from Multi-scale ReFinement of Super-regions (SMURFS), which not only obtains state of the art superpixels, but can also be applied hierarchically to form what we call  $n$ -th order super-regions. In essence, starting from a uniformly distributed set of super-regions, the algorithm iteratively alternates graph-based split and merge optimization schemes which yield superpixels (1st order super-regions) that better represent the image. We define a super-region hierarchy forming the level  $i$  by grouping elements of the level  $i - 1$ . Denoting the pixel grid as level  $i = 0$ , superpixels (level  $i = 1$ ) are formed by grouping similar adjacent pixels while *supersegments* (level  $i = 2$ ) are formed of multiple superpixels. To be able to better represent the image, we alternate optimization schemes at both level  $i = 1$  and  $i = 2$  with the aim of refining both superpixels and super-segments simultaneously. The split step is performed over the pixel grid to separate large super-segments into different smaller superpixels. This step is fully parallelizable as every region is split independently, and produces superpixels that better capture local information of the super-segments. The merging process, conversely,

is performed over the superpixel graph to create *supersegments* with the aim of better capturing global image features. This iterative two-scale procedure refines the super-region boundaries of the image without shape or boundary initialization constraints, present in most of state of the art superpixels. Results show state of the art Achievable Segmentation Accuracy (ASA) in the Berkeley Segmentation dataset (BSD500) [1].



**Figure 2:** ASA comparison on the BSD500 dataset.

- [1] D. Martin, C. Fowlkes, D. Tal, and J. Malik. A database of human segmented natural images and its application to evaluating segmentation algorithms and measuring ecological statistics. In *Proc. 8th Int'l Conf. Computer Vision*, volume 2, pages 416–423, July 2001.