

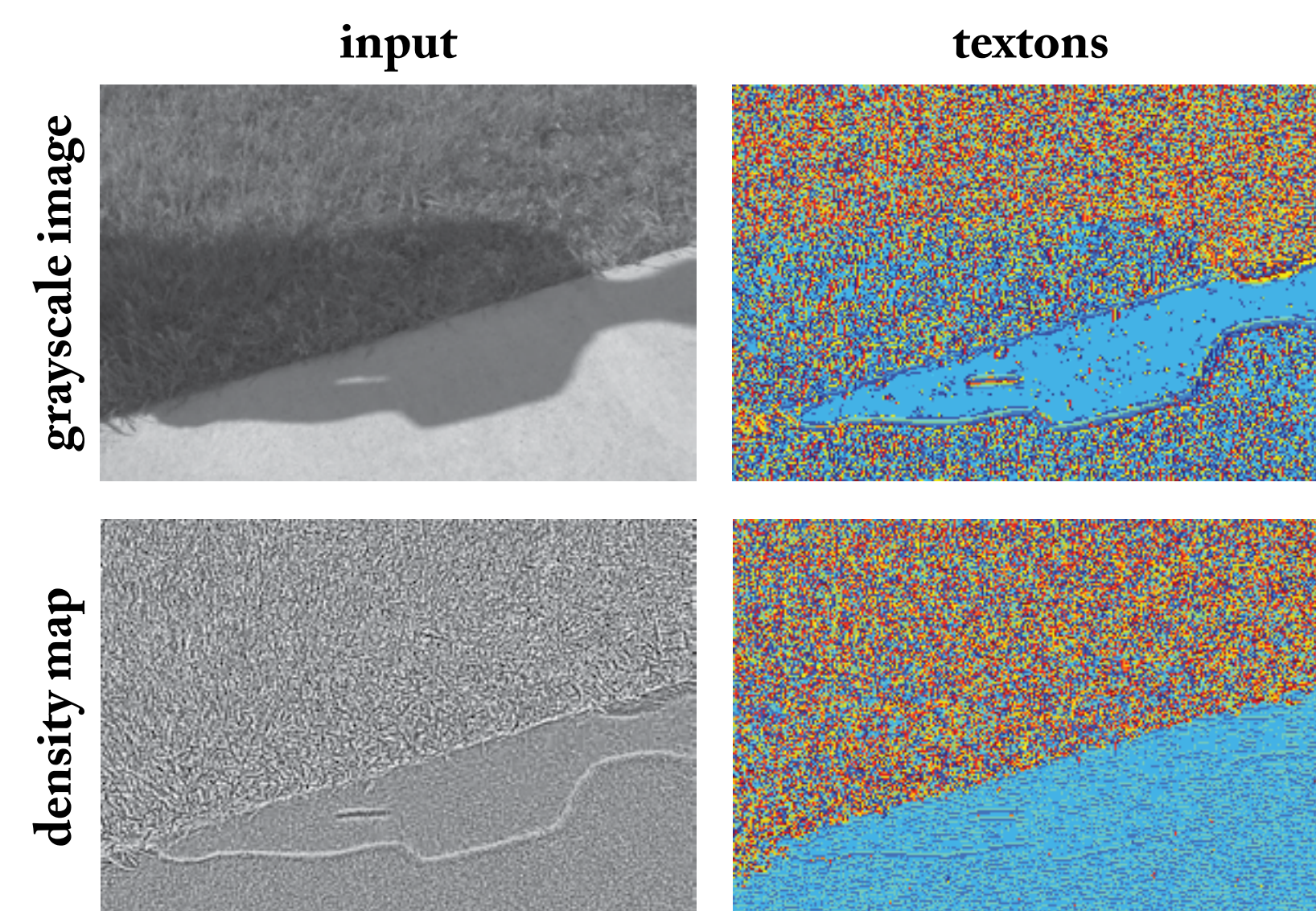
# Shadow free segmentation in still images using local density measure

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## Image Density Map

**Problem:**  
classical filterbank based texture descriptors are not robust to shadows

**Solution:**  
preprocess images using local density measure



## Formulation

• Define a measure function on the image over radius  $r$

$$\mu(\mathbf{x}, r) = \sum_{\|\mathbf{y}-\mathbf{x}\| \leq r} I(\mathbf{y})$$

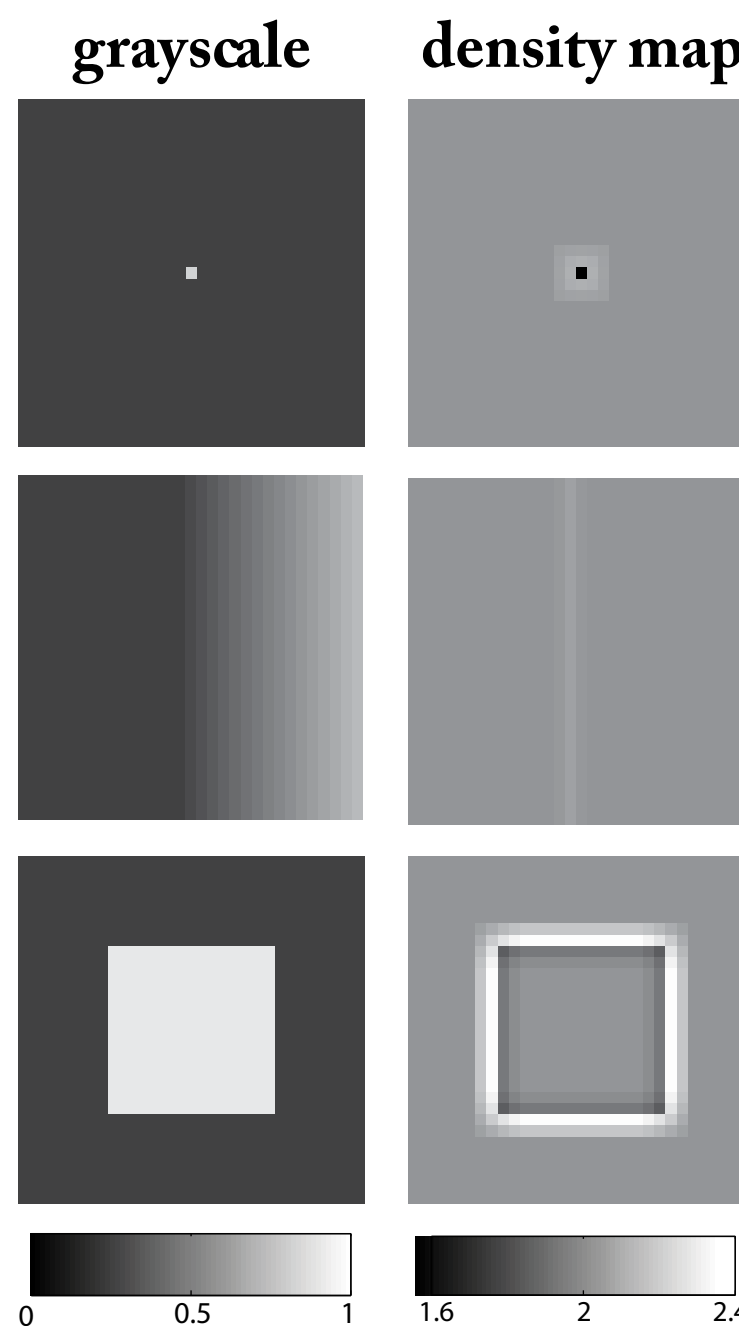
• Hypothesize that it varies as an exponential of  $r$

$$\mu(\mathbf{x}, r) = kr^{d(\mathbf{x})}$$

$$\log(\mu(\mathbf{x}, r)) = \log k + d(\mathbf{x}) \log r$$

$$d(\mathbf{x}) = \lim_{r \rightarrow 0} \frac{\log(\mu(\mathbf{x}, r))}{\log r}$$

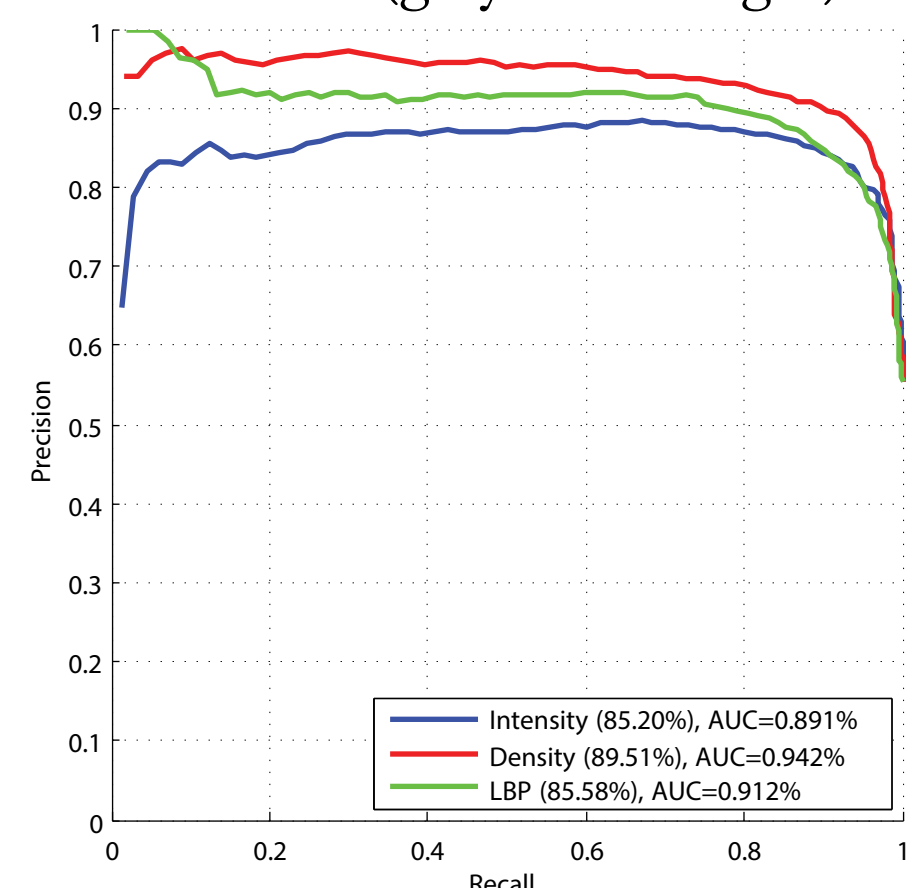
- $d(\mathbf{x})$  is the local density function of image  $I(\mathbf{x})$  at pt.  $\mathbf{x}$
- Intuitively measures the degree of regularity of intensity variation in a local neighbourhood around  $\mathbf{x}$
- Important properties:
  - Preserves texture details
  - Invariant to multiplicative illumination changes



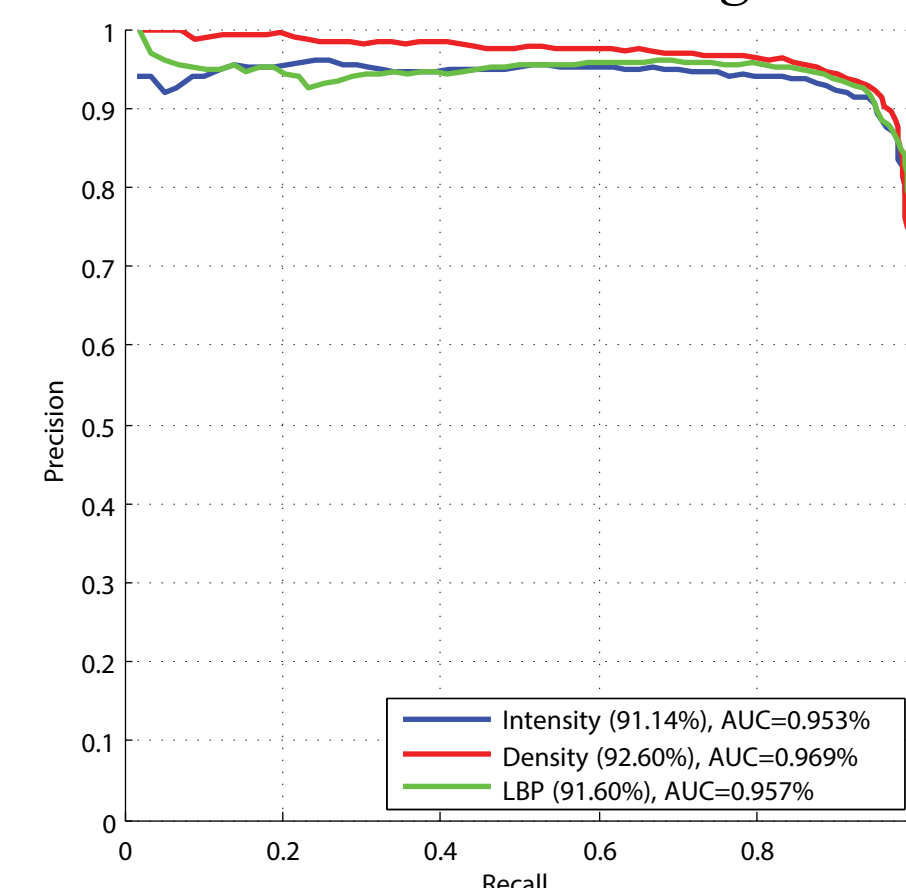
## Application to shadow boundary detection

- Goal: detect shadow boundaries in images
- For each edge pixel compare appearance features on both sides of the edge
- Features:
  - intensity ratios
  - RGB colour channel ratios
  - textons on intensity | textons on density | Local Binary Patterns
- Evaluate on shadow image dataset by Zhu et al
- Two experiments
  - Grayscale images (intensity and texture features only)
  - Colour images (intensity, texture and colour features)

PR curve (grayscale images)



PR curve (colour images)



## Shadow free segmentation

• Graph cut image segmentation formulation

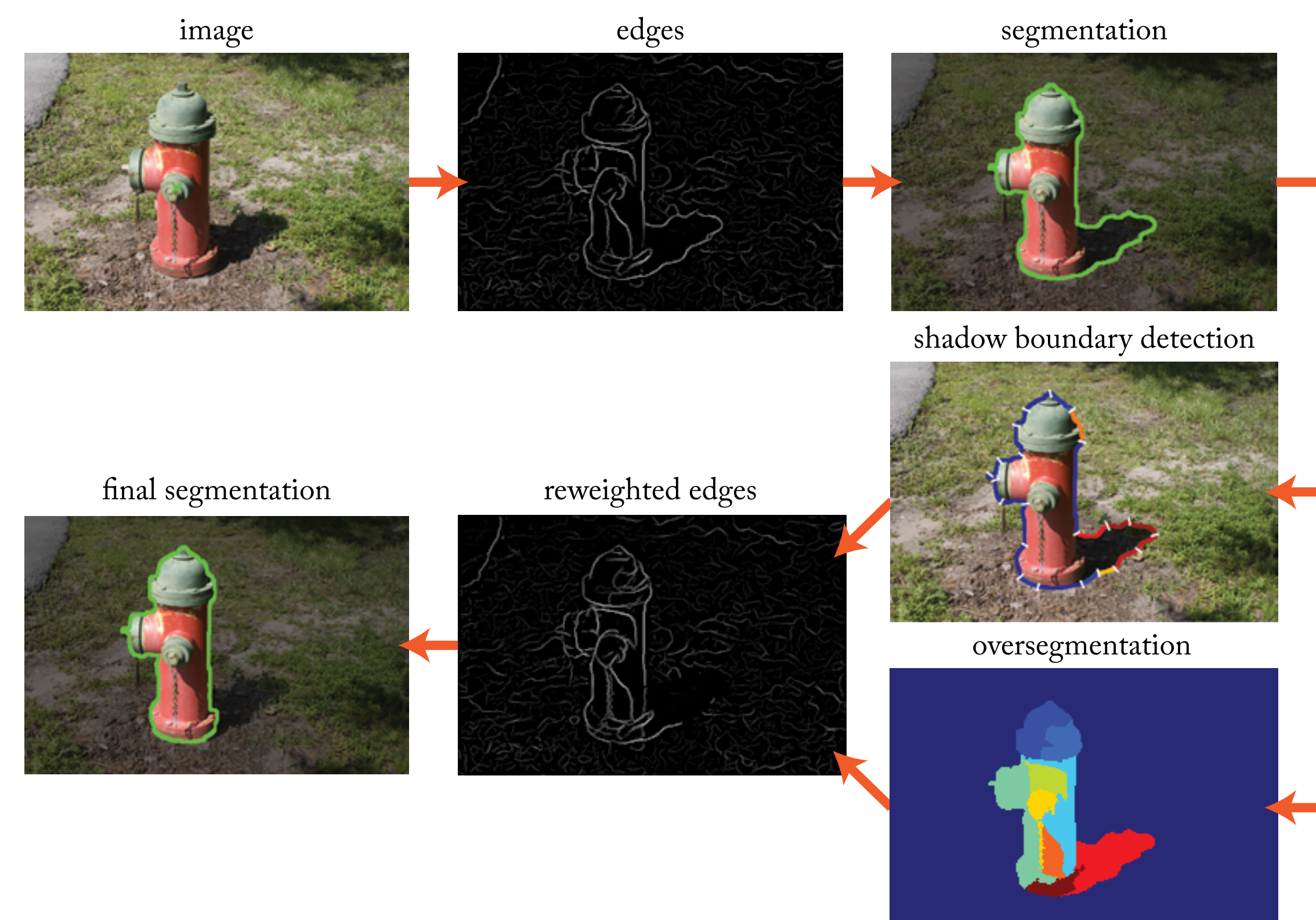
$$E(f) = \sum_{p \in \mathcal{P}} D_p(f_p) + \lambda \sum_{\{p, q\} \in \mathcal{N}} V_{p, q} \cdot \delta(f_p \neq f_q)$$

- Shadows change the colour model of the shadowed region
  - use an illumination invariant colourspace ( $L^*a^*b^*$ )
- Shadows affect the binary weights by
  - introducing string gradients at the boundary of shadow region
  - reducing gradients in shadowed regions
- Two most common segmentation mistakes
  - undersegmentation
  - oversegmentation



## Our approach

- Observation: wrong segmentation boundary always goes along shadow boundaries
- Our approach:
  1. segment
  2. modify binary weights
    - attenuate shadow boundary edges
    - reinforce internal edges
  3. resegment



## Evaluation

- Dataset of 52 outdoor images
- Manually labelled figure ground segmentations
- Baseline: segmentation without shadow removal
- Test our method with three different shadow detectors
  - intensity textons
  - density textons
  - Local Binary Patterns

Algorithm	F-measure
Baseline	0.77±0.033
Intensity textons	0.80±0.052
Density Textons	0.84±0.045
LBP	0.81±0.055

## Results

