

# Estimating the Ground Truth From Multiple Individual Segmentations Incorporating Prior Pattern Analysis with Application to Skin Lesion Segmentation

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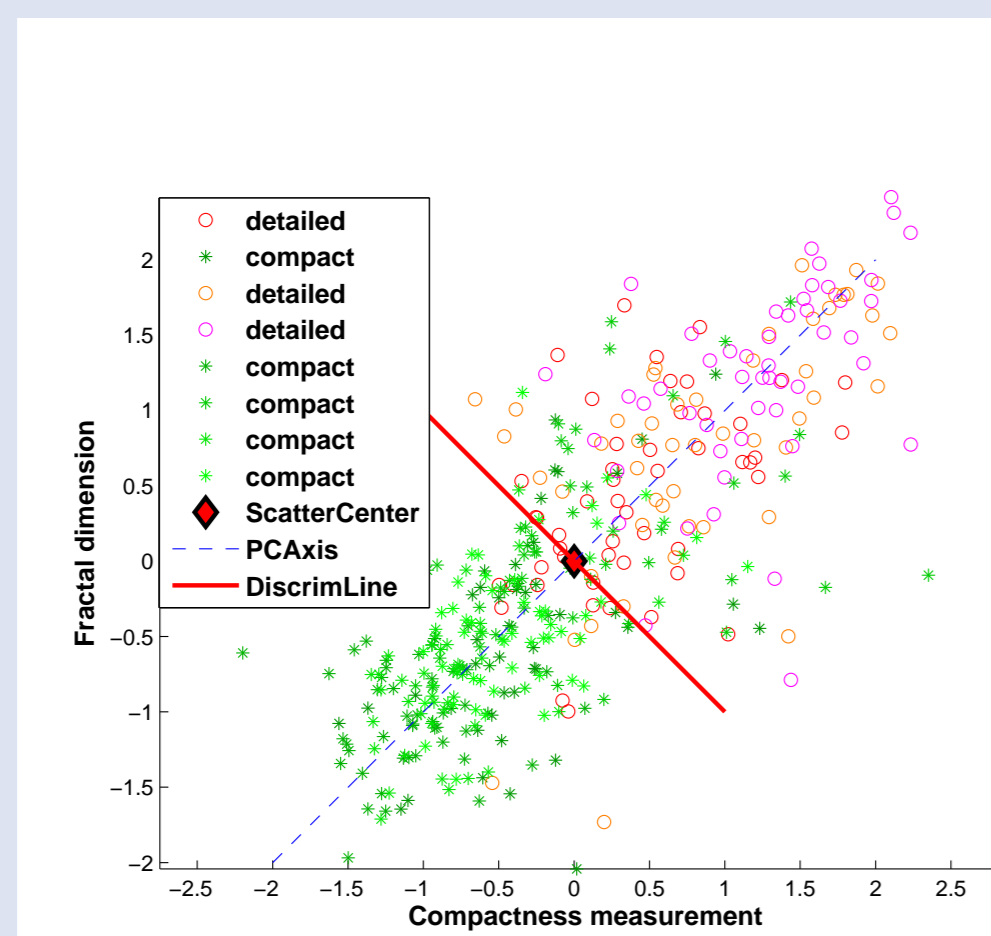
## Claim

A proper lesion boundary ground truth estimation approach should take into account and compensate for the inter-rater variation.

## Manual Segmentation Patterns

Question: Are there different segmentation patterns?

- **Subject:** Inter-rate variation
- **Object:** 50 lesion images / 8 dermatologists
- **Pattern analysis features:** Compactness Measurement (CM) and Fractal Dimension (FD)
- **Result:** two segmentation patterns (**detailed** and **compact**) exist because of different segmentation policies



The scatter plot of FD and CM



Manual segmentations categorized using the above analysis

## Ground Truth Estimation Methods Solved using Level-set approach

- Maximize the *a posteriori* (MAP) probability based energy function (LSML)

$$E_{LSML} = - \sum_n \sum_{x \in \Omega_n} \log p(T(x) | D_{\{1,2,\dots,J\}}(x)) \quad (1)$$

(2)

- Segmentation pattern information based energy function

$$E_{shape} = \int_{\Omega} [T(x) - SPM(x)]^2 dx \quad (3)$$

(4)

- The ground truth estimation energy function (LSMLP)

$$E = E_{LSML} + E_{shape}. \quad (5)$$

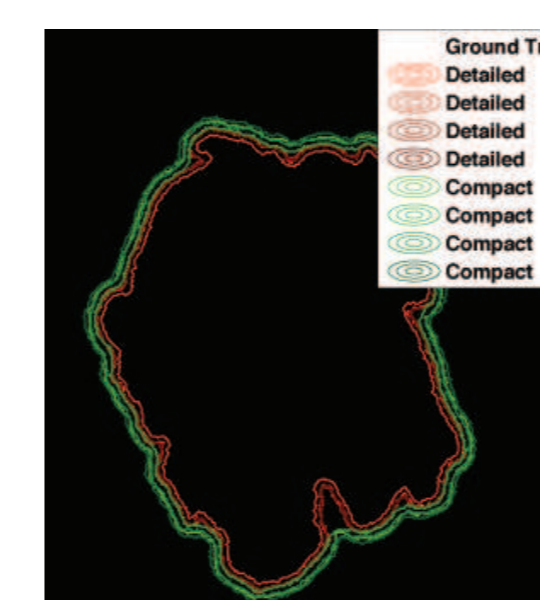
- The ground truth calculation equation: maximizing the energy function

$$\frac{\partial \phi}{\partial t} = - \frac{\partial E(\phi)}{\partial \phi} = \delta(\phi) \left( \log \frac{W}{V} + \gamma \times (2 \times SPM(x) - 1) \right). \quad (6)$$

- **Notations:**  $D_{\{1,2,\dots,J\}}(x)$ : manual segmentations at pixel position  $x$ ;  $T(x)$ : the estimated ground truth;  $SPM$ : shape prior model learnt from manual segmentation pattern (see paper for details);  $\gamma$  weights the importance of the shape prior energy;  $W$  and  $V$  are the joint conditional probability that pixel  $x$  belongs to the lesion and skin, respectively.

## Experiments

- **Evaluation metrics:** XOR and FOM
- **Generate synthetic data for testing**



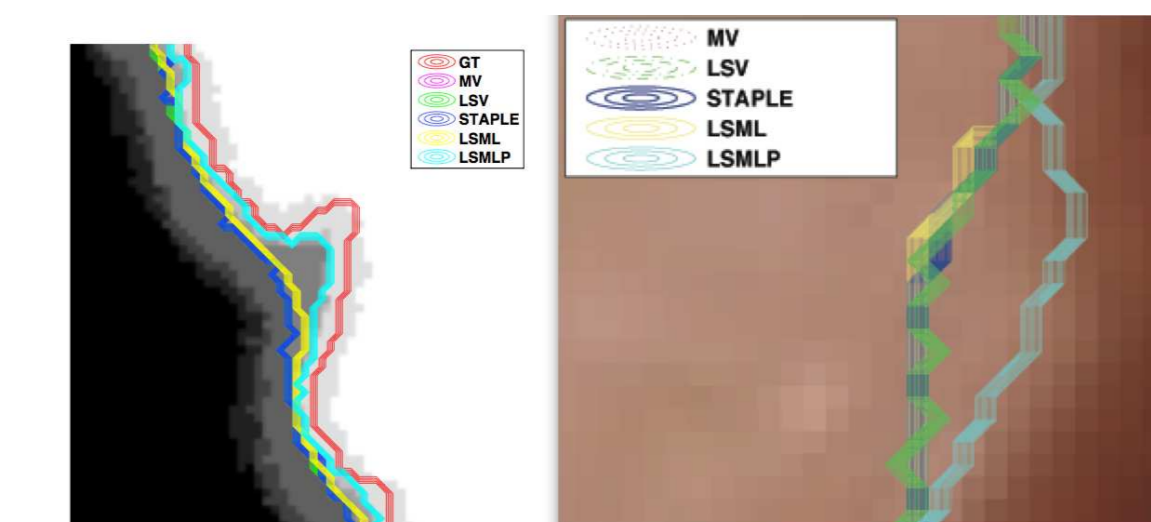
The synthetic segmentations

- **Comparisons:** LSMLP has the best performance.

Metrics	Methods				
	MV	LSV	STAPLE	LSML	LSMLP
<b>XOR (%)</b>	3.8409	3.8409	3.7212	3.2733	2.1615
<b>FOM (%)</b>	8.9026	8.9026	10.6596	13.1484	26.7412
<b>Sensitivity</b>	1.0000	1.0000	1.0000	1.0000	1.0000
<b>Specificity</b>	0.9709	0.9709	0.9719	0.9754	0.9839

The performance of different ground truth estimation methods

**Notations:** MV: majority vote method; LSV: variation minimization based approach (which is proved to be identical to MV, see details in paper); STAPLE [Warfield2004].



Comparison on synthetic image (Left) and real image (Right)

## Conclusion

- Experiments on both synthetic and real data show that segmentation style prior information helps to find a more accurate estimate of the ground truth.
- LSMLP uses this prior information to produce a ground truth that has smaller error.

## Reference

Simon K Warfield, Kelly H Zou and William M Wells. Simultaneous truth and performance level estimation (staple): an algorithm for the validation of image segmentation. IEEE Transactions on Medical Imaging, 23(7):903–921, 2004.