



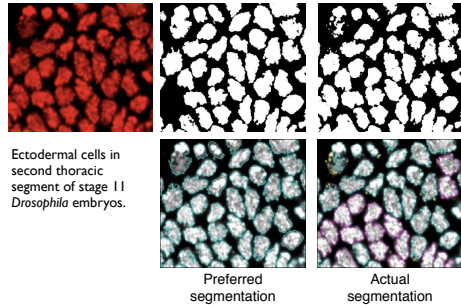
SEGMENTATION OF NUCLEI IN CONFOCAL IMAGE STACKS USING PERFORMANCE BASED THRESHOLDING

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Problem

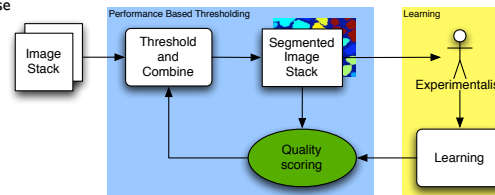


Challenge

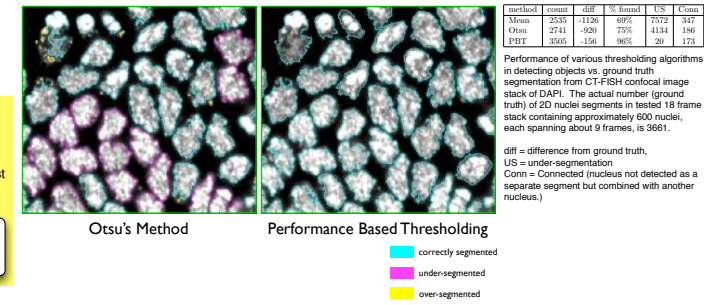
- Tuning required for computer vision algorithms (for both initial use and as experimental conditions change)
- Tuning requires computer vision expertise
- experimentalist ≠ computer vision expert (typically neither side wants =)

Our Approach

- use machine learning to *mimic* biologist
- intuitive interface engages biologist in training machine



Results



Performance Based Thresholding

```

1: for all framea, in stack do
2:   segmentsa = {}
3:   for all thresholds t from low to high do
4:     B = framea > t
5:     CE = 2D connected elements in B.
6:     for all cej in CE do
7:       p = P(cej)
8:       if p ids "good" then
9:         mask (remove) cej from framea.
10:        add cej to segmentsa.
11:       end if
12:     end for
13:   end for
14:   Connect segments between segmentsa and segmentsa-1
15: end for
  
```

Pseudocode for Performance Based Thresholding (PBT)

Iteration 2: frame_{a+1}, frame_a, segments_a

Iteration 31: frame_{a+1}, frame_a, segments_a

Iteration 98: frame_{a+1}, frame_a, segments_a

Learning a Shape Recognizer

Experimentalist correct labeling

Boosting

Alternating Decision Tree (Quality scoring)

Extract features

object + label

Boosting Details

A weak learner

The Boosting Process

AdaBoost

$$F_0(x) = 0$$

$$\text{for } t = 1..T$$

$$w'_t = \exp(-y_i F_{t-1}(x_i))$$

Get h_t from weak-learner

$$\alpha_t = \frac{1}{2} \ln \left(\frac{\sum_{i: h_t(x_i) = y_i} w'_t}{\sum_{i: h_t(x_i) \neq y_i} w'_t} \right)$$

$$F_{t+1} = F_t + \alpha_t h_t$$

The weak requirement: $|\sum_{i=1}^n y_i \hat{y}_i w_i| > \gamma > 0$

$$F_T(x) = \alpha_1 h_1(x) + \alpha_2 h_2(x) + \dots + \alpha_T h_T(x)$$

$$f_T(x) = \text{sign}(F_T(x))$$

[Freund, Schapire 1997]