

# Fast semi-supervised segmentation of in situ tree color images

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# Context

- Regular diagnosis of urban trees to assess their **health** and... **potential risk**
- By visually looking to **discontinuities** in shape and density
- Based on digital photographs taken **in situ**



*Winn, M.F., Araman, P.A., Lee, S.M. "Urban Crowns: an assessment and monitoring tool for urban trees", Gen. Tech. Rep., U.S. Dept. of Agriculture, Forest Service, (2011)*

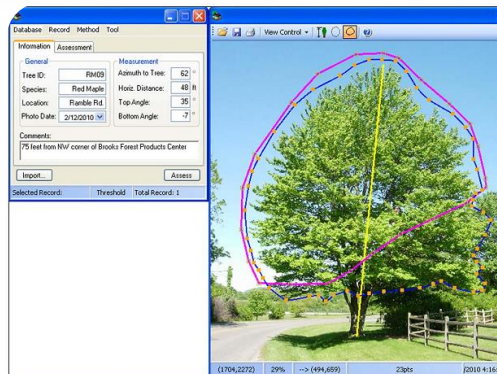


Figure 5—Screen capture of the UrbanCrown program showing the Data Control window on the left and the Tree Image window on the right.



Figure 9—Post-assessment results for the sugar maple example.

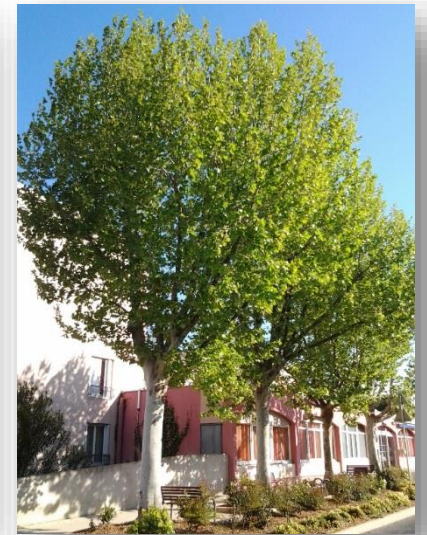
# Context

→ Segmenting a tree in natural color images

- Non uniform background

- Lighting problems
- Occulting objects

- Overlap with other trees



→ On a standard mobile device

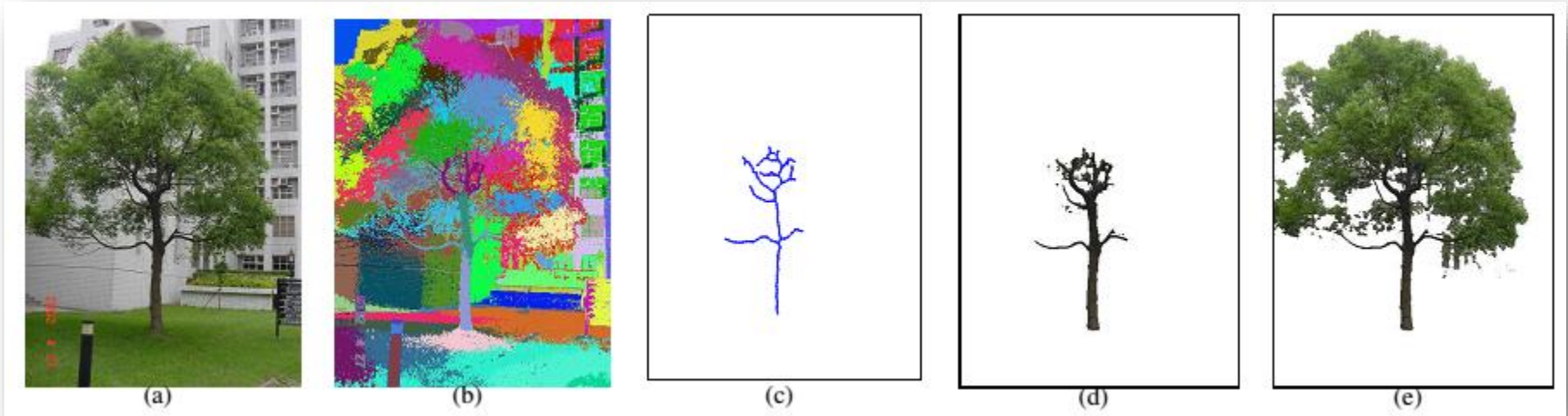




# Some solutions?

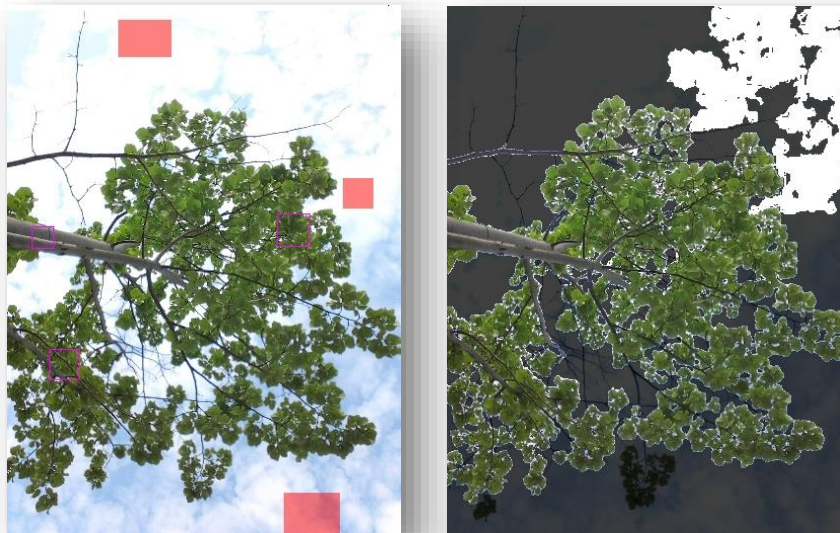
- By making some hypotheses about the tree structure

Too complex scene?



*Teng, C.H., Chen Y.S., Hsu, W.H. "Tree segmentation from an Image". In: IAPR Machine Vision Appl., 59-63 (2005)*

- By asking the operator to tag different tree parts



*Friedland, G., Jantz, K., Rojas, R.: "SIOX: simple interactive object extraction in still images". In: 7th IEEE International Symposium on Multimedia (2005)*

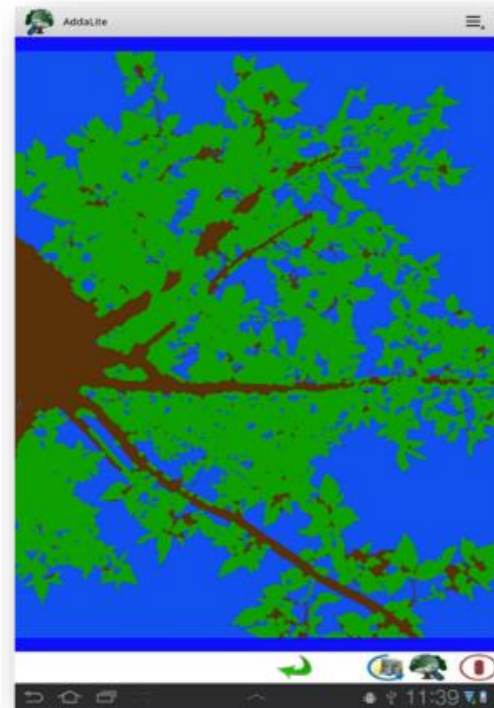
Too heterogeneous background?

# Proposed method

1. Reduce the color dynamic range with a **quick algorithm**.
2. The operator quickly and roughly tags some parts of the image which become **learning areas (LA)** for the labeling process...  
  
...which will then classify **all the pixels** of the image.
3. A post-processing will **emphasize** narrow and thin structures.

→ **3 labels :**

- Leaves
- Wood
- Background

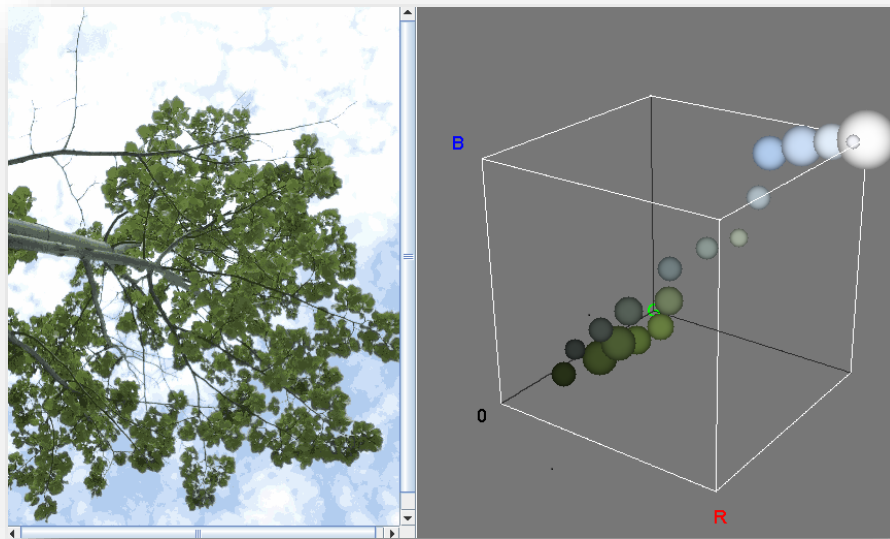


# 1. Reduction of the color image dynamic range

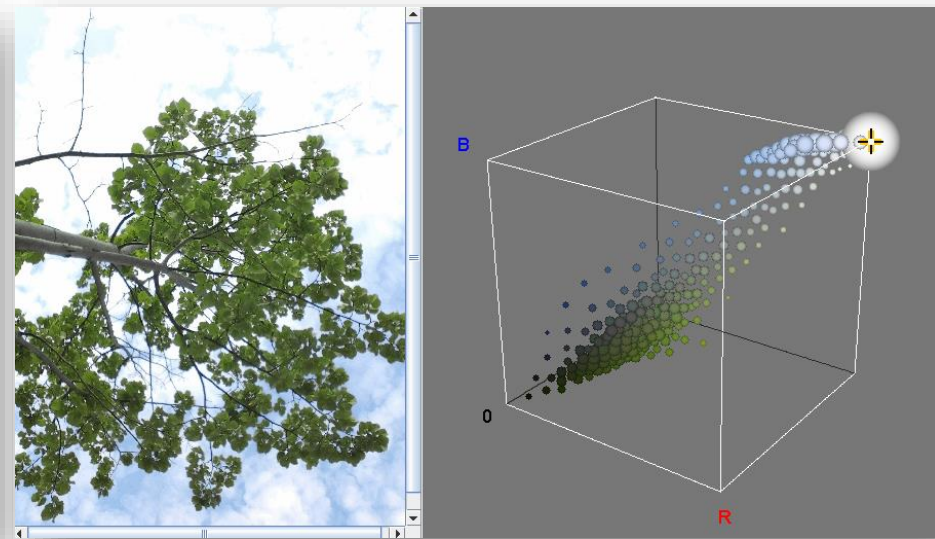
From native 24-bit to 8-bit to reduce complexity

*Wu, X. "Color quantization by dynamic programming and principal analysis". In: ACM Transactions on Graphics. 11(4) : 348-372 (1992)*

- PCA in the color space
- Recursive decomposition to reach a given number of colors



20 colors



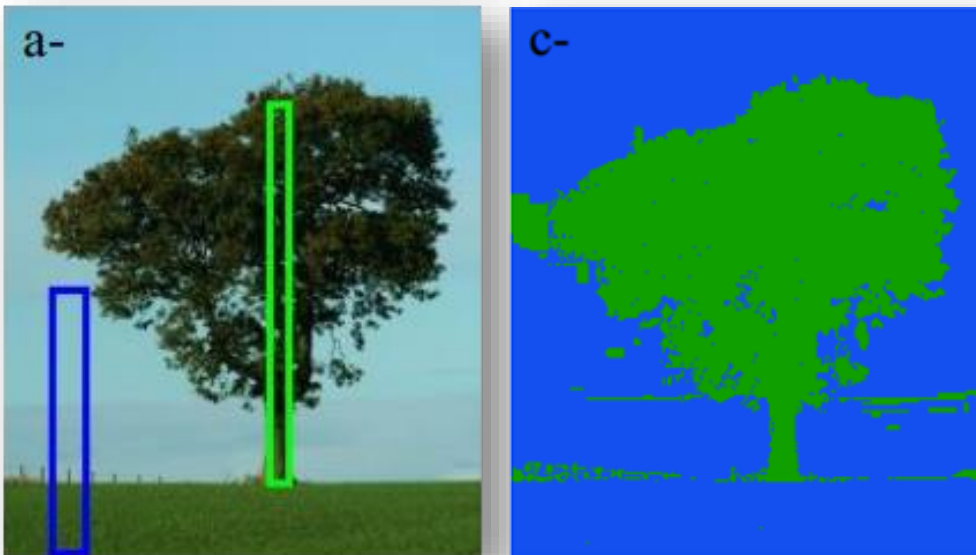
200 colors

(example in RGB color space)

## 2. Labeling of the image

**Strategy A** : from the most to the less homogeneous LA

1. Sort all the LAs w.r.t. to their variance
2. Along the LAs
  - i. Assign the corresponding label to all the colors belonging to the LA
  - ii. Remove all these labelled colors to the remaining LAs.
3. Compute new variances of remaining LAs and reorder

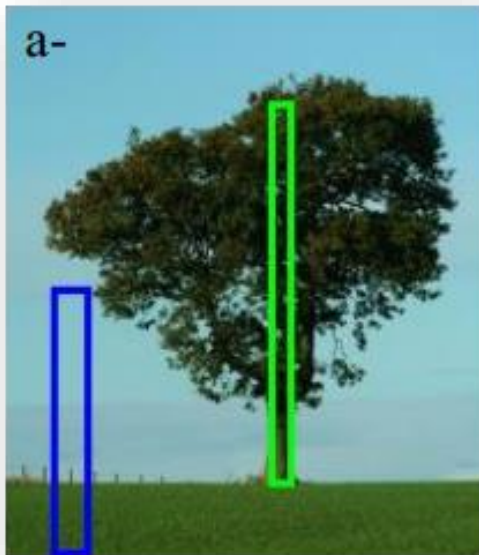


Aberrant labelling caused by the heterogeneity of LA?

## 2. Labeling of the image

**Strategy B** : fragmenting heterogeneous LAs

0. *Fragment all the LAs in subLAs by a k-means clustering to be under a maximal variance*
1. *Sort all the subLAs w.r.t. to their variance*
2. *Along the subLAs*
  - For each color of the subLA, compute the distance to the current and already processed subLAs.*
  - Assign the label corresponding to the subLA with minimal distance*
  - Remove all these labelled colors to the remaining subLAs*
3. *Compute new variances of remaining subLAs and reorder*

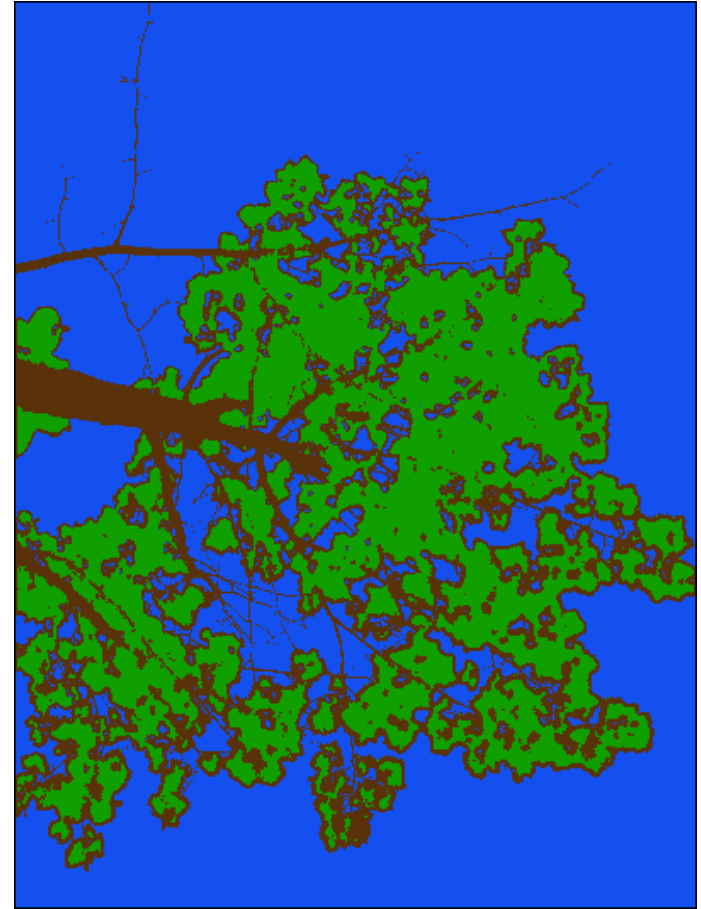


Better results?



# Stage 3: post-processing to emphasize narrow and thin structures

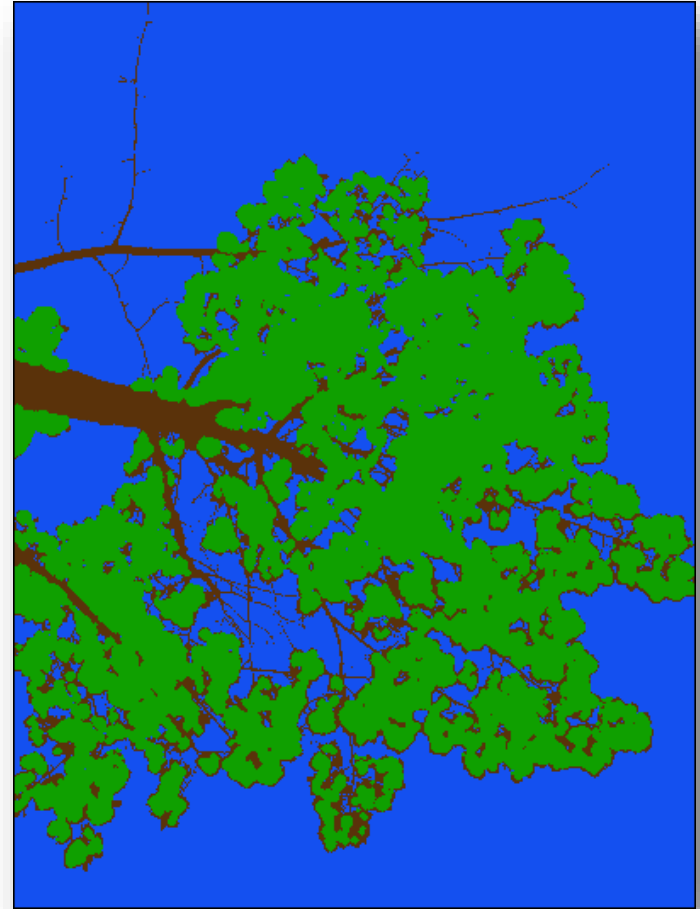
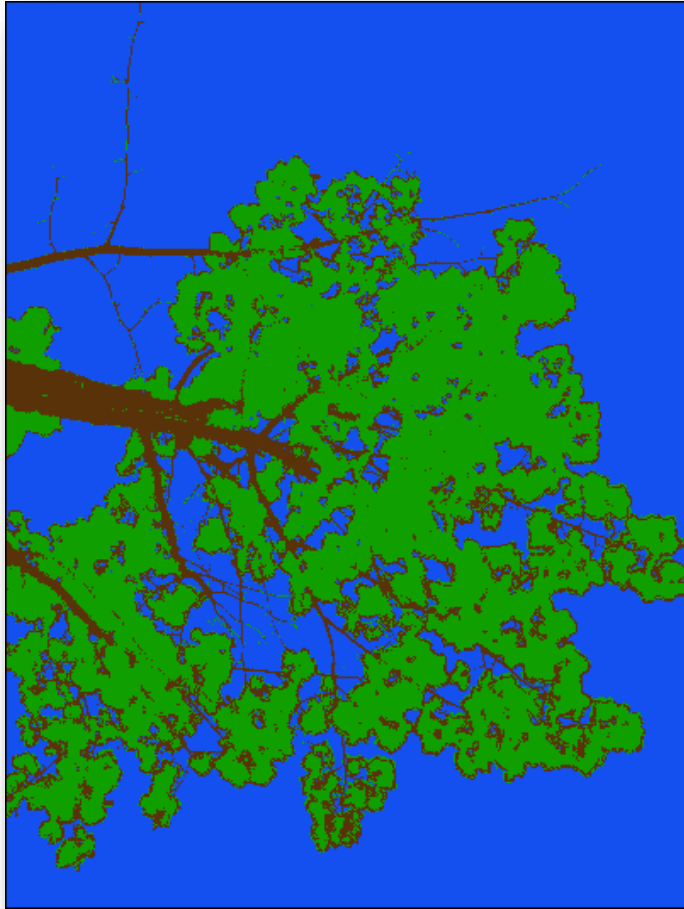
- Conditional dilations of “wood” at the expense of “leaves” in order to strengthen branches



$$W_{i+1} = \{p \in W_i\} \cup \{p \in L_i \mid S_{p,w} \not\subset L_i\} \text{ and } L_{i+1} = \{p \in L_i \mid S_{p,w} \subset L_i\} \text{ with } W_0 = W, L_0 = L$$

# Stage 3: post-processing to emphasize narrow and thin structures

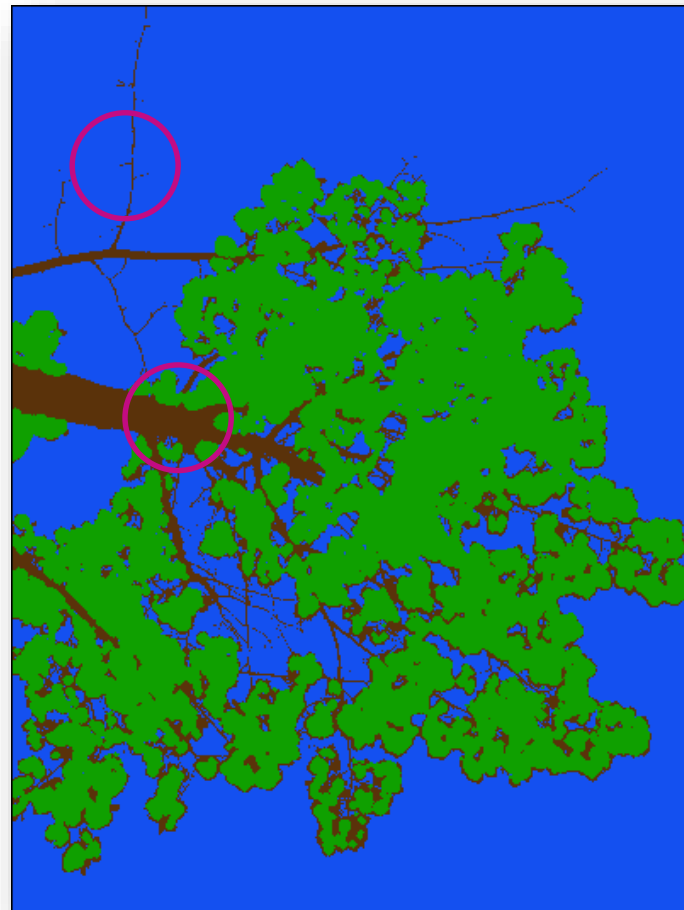
- Conditional dilations of “wood” at the expense of “leaves” in order to strengthen branches
- Opposite conditional dilations of “leaves” at the expense of “wood”.



$$W_{i+1} = \{p \in W_i\} \cup \{p \in L_i \mid S_{p,w} \not\subset L_i\} \text{ and } L_{i+1} = \{p \in L_i \mid S_{p,w} \subset L_i\}$$
$$L_{i+1} = \{p \in L_i\} \cup \{q \in W_i \mid q \in S_{p,w}, p \in L_i\} \text{ and } W_{i+1} = \{p \in W_i \mid p \notin S_{q,w}, q \in L_i\}$$

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# Discussion

- Sensitivity to the learning areas?

  - BLA= 3 large heterogeneous LAs

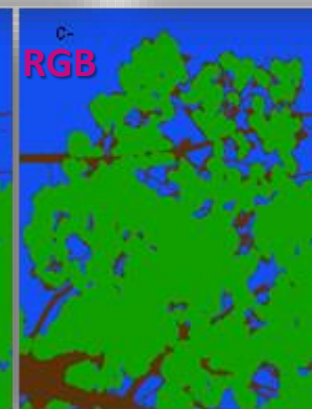
  - SLA= 9 small homogeneous LAs

    - Strategy B for using BLA



- Optimal color system?

  - L\*a\*b\* better



- Other parameters?

  - ....found experimentally...

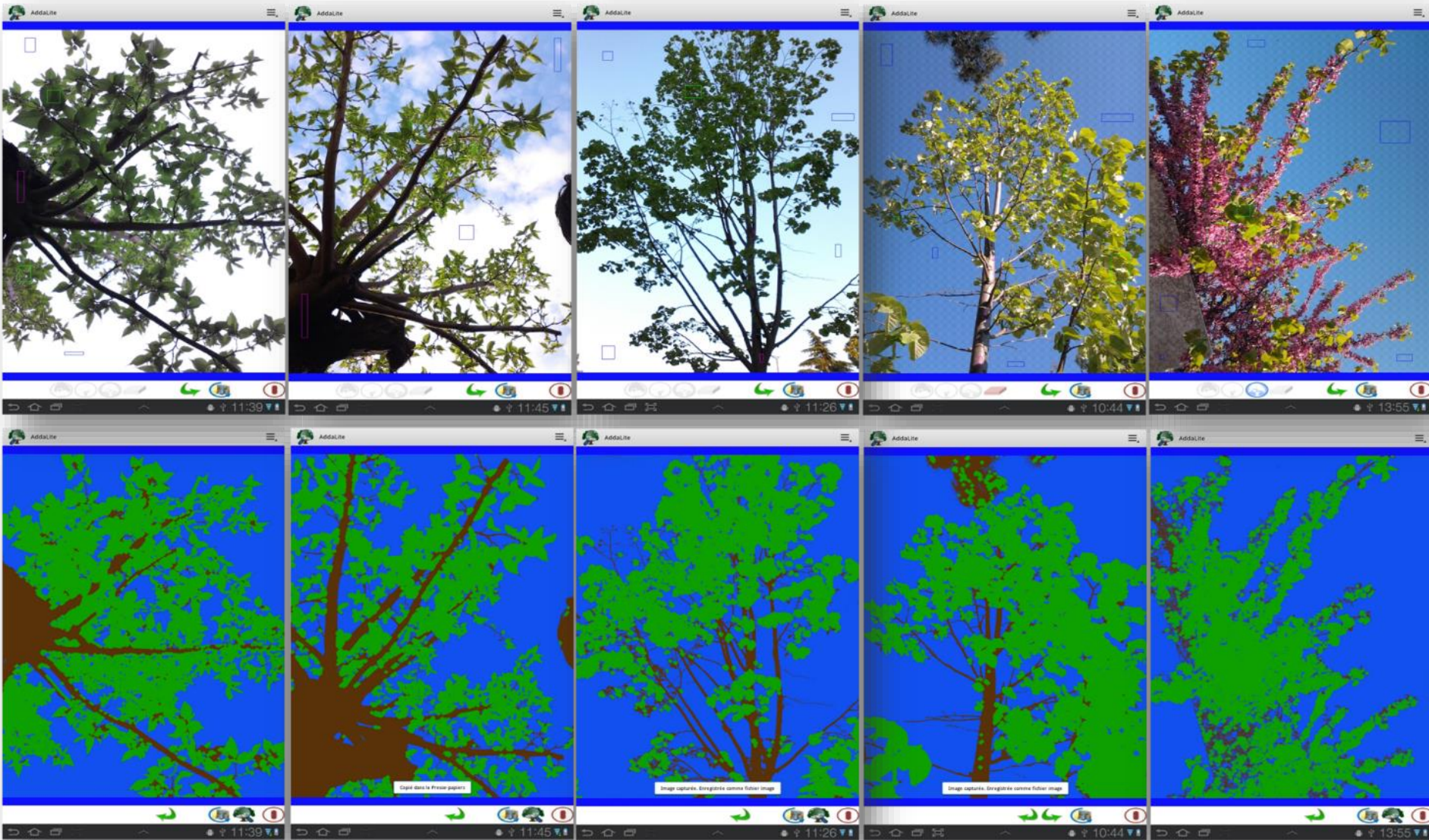
    - Step 1: number of colors=200

    - Step 2, strategy B, maximal variance for fragmentation=10

    - Step 3: number of conditional dilations=3



# Experimental results

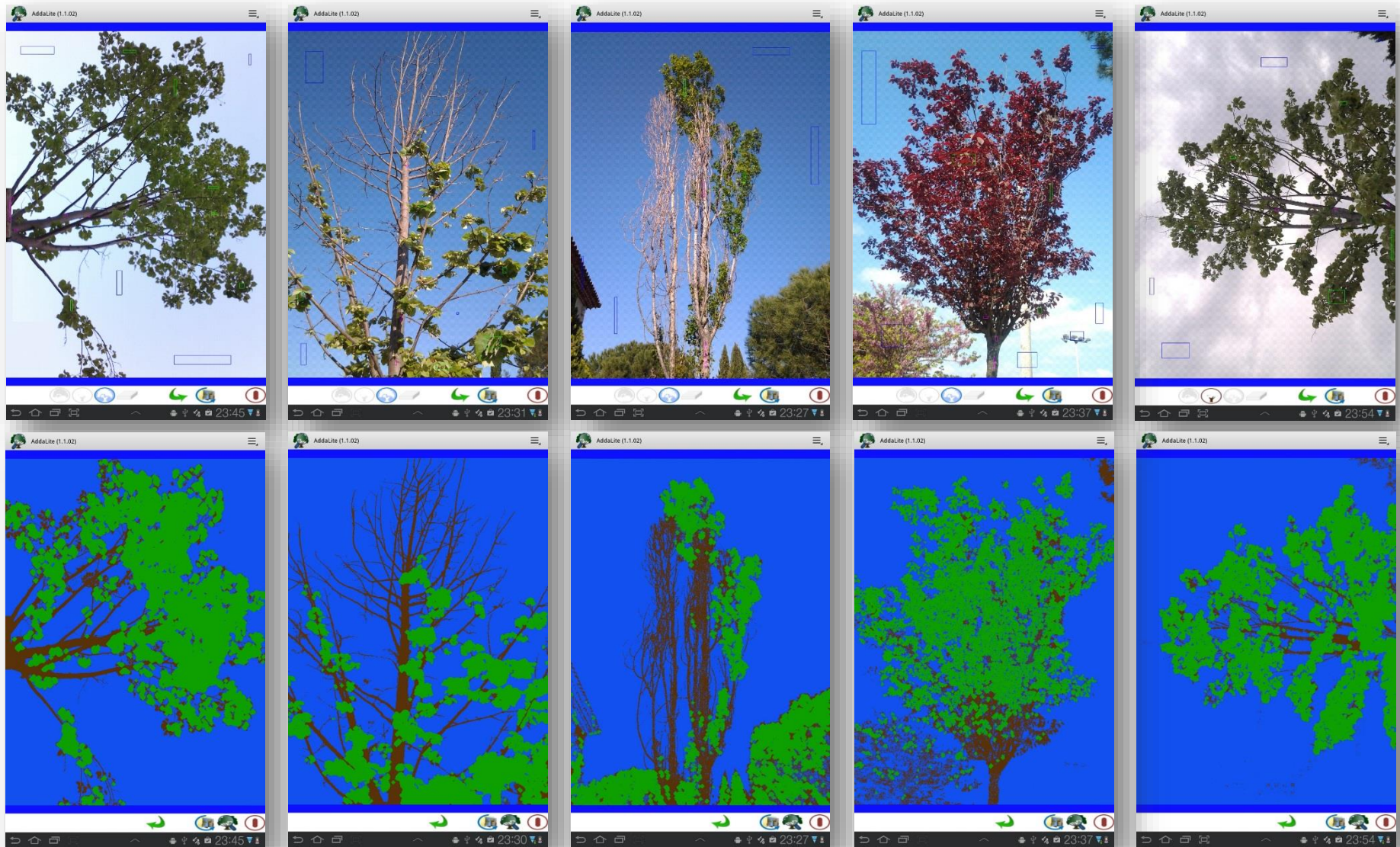


Samsung Galaxy 2 tablet (1 GHz Nvidia Tegra)

35 s for a 2408 x 1536 pixel image



# Experimental results

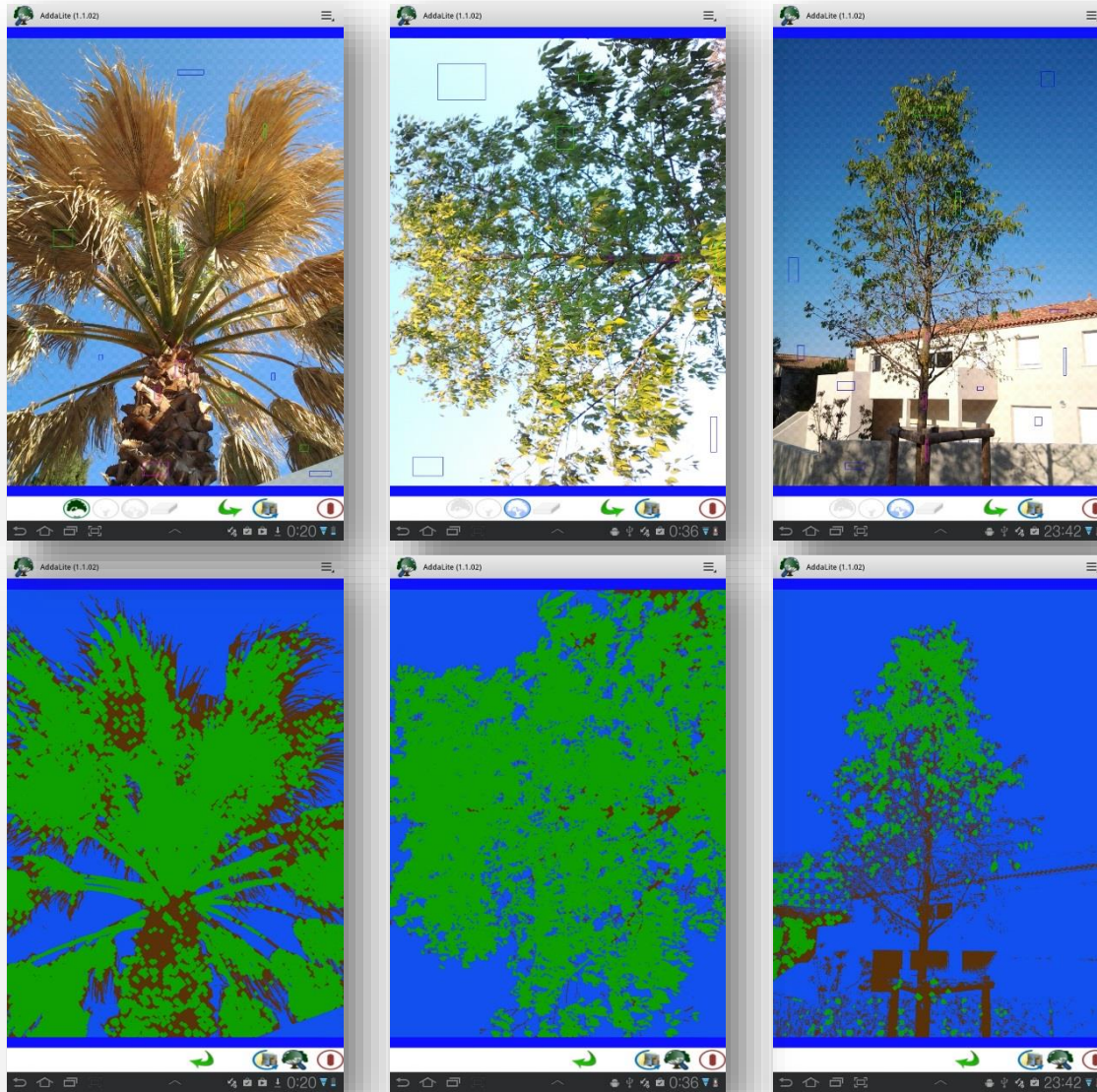


Samsung Galaxy 2 tablet (1 GHz Nvidia Tegra)

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# Experimental results

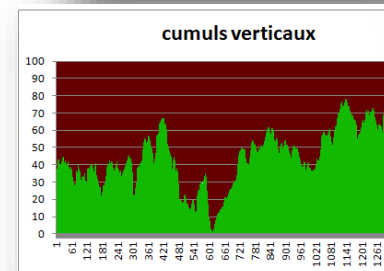
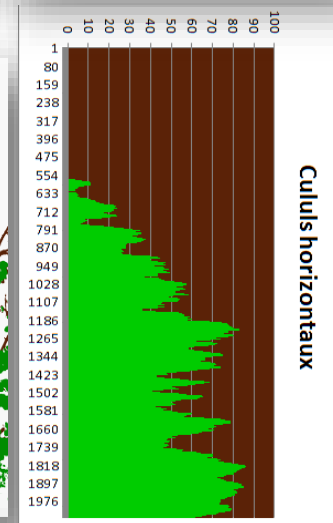
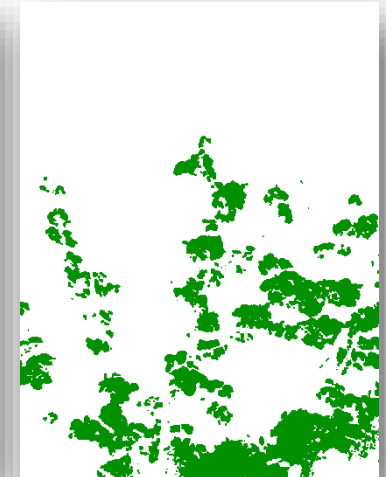


Samsung Galaxy 2 tablet (1 GHz Nvidia Tegra)

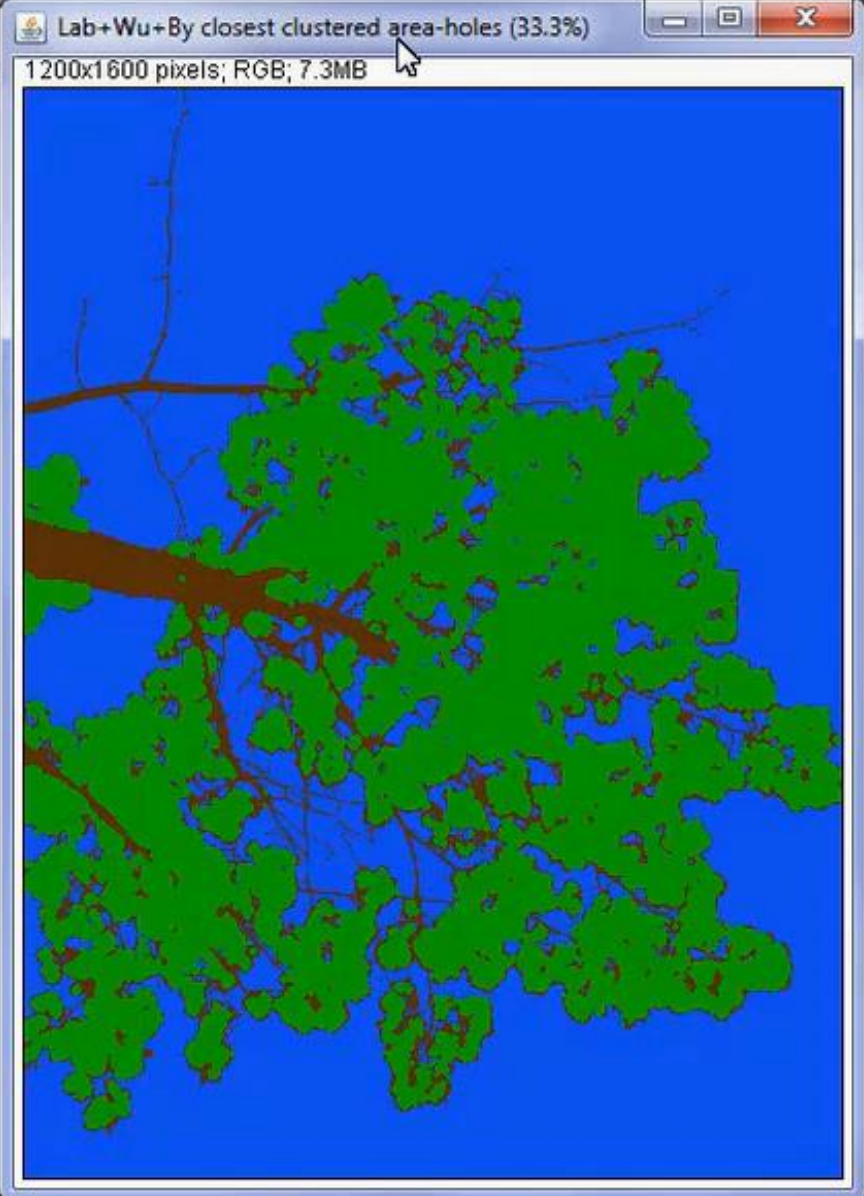
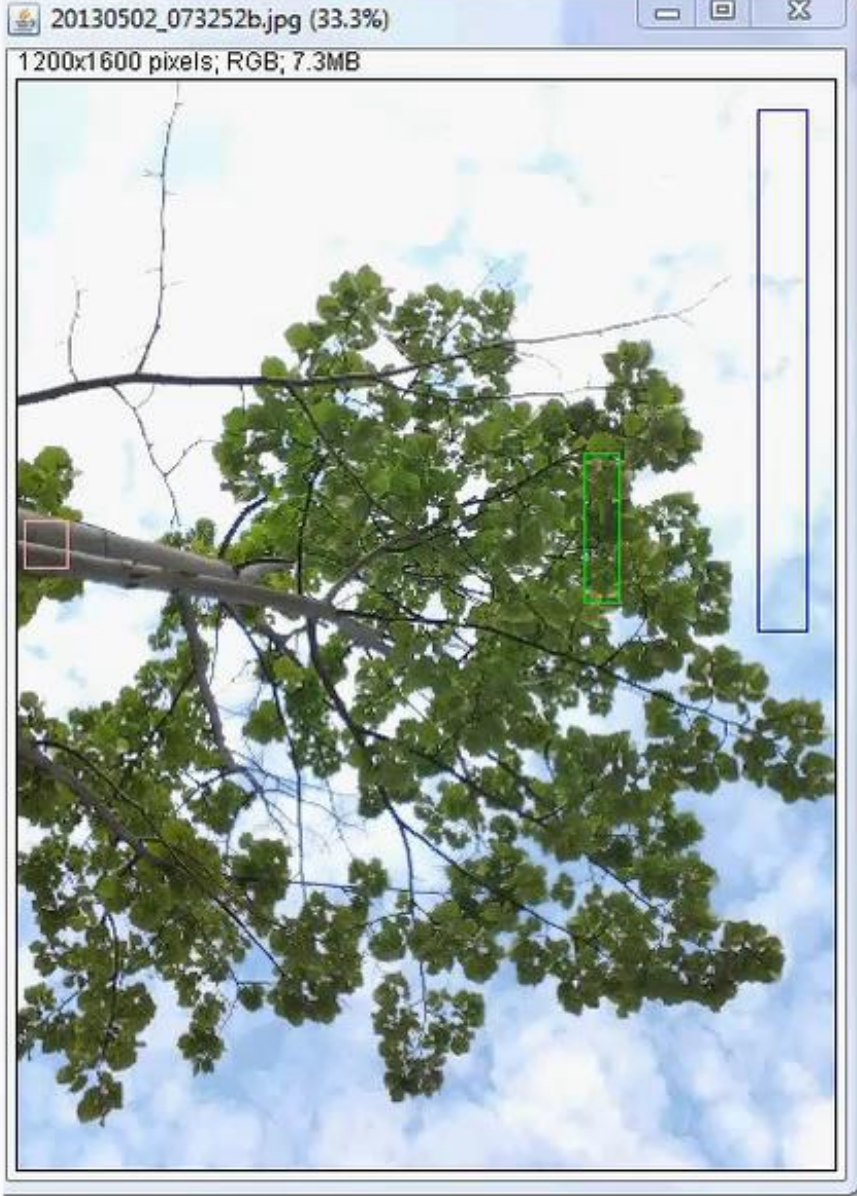
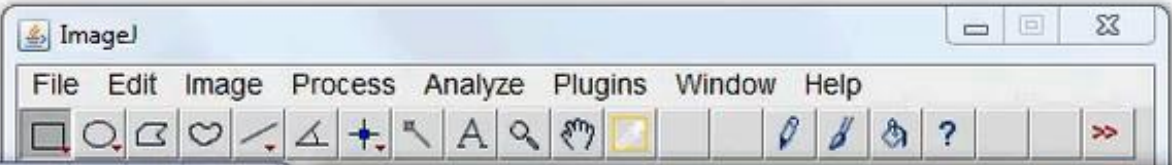
35 s for a 2408 x 1536 pixel image

# Future work

- Computing tree health parameters:
  - ✓ crown transparency
  - ✓ proportion of wood, leaves
  - ✓ regularity of crown shape
- Improve implementation and performance
- Improve ergonomony for a practical application by a non-specialized operator
- Assess the results w.r.t. analyses by experts







Thank you for your attention !

