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Leaf Identification Using Image Processing and Automatic Pattern Recognition

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Sample Images



Ginkgo



Tulip Tree



Sweet Gum



Sugar Maple



Fig Tree



White Birch



White Oak



Sassafras Tree



Shagbark Hickory



Flowering Dogwood

Leaf Identification Using Image Processing and Auto Pattern Recognition

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Introduction: We are fearfully and wonderfully made in the image of God. This is evidenced in how the human brain is able to process and make decisions based on what it sees. In this project we designed a system that identifies and classifies leaves just like a human would.

Problem: In order to distinguish between different leaves, a certain set of criteria had to be established. We needed to use as few leaf parameters as possible in order to correctly identify each leaf.

Materials and Methods: Using Matlab as a base, we utilized various image processing techniques to create the desired criteria. In order to accurately identify each leaf, we used a combination of four parameters. Rectangularity, convex area, perimeter, and the spectral peak location made up our identification criterion.

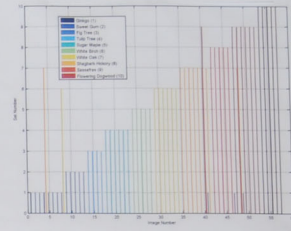
Rectangularity: This multiplies the longest possible combination of length and width of the leaf. It then takes this area and divides it by the actual leaf area. This tells us if how closely a leaf reflects the shape of a rectangle.

Perimeter: This measures the pixel distance around a leaf.

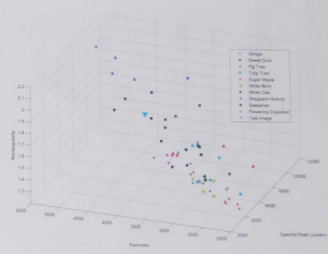
Convex Area: We first fill in all concavities native to each leaf so that we are left with a convex version of the leaf. We then measure the area of this convex leaf and divide it by the original leaf area. This gives us a ratio of convex leaf area to the original leaf area.

Spectral Peak Location: Visualize placing a point at the center of a leaf and then measuring the distance to the edge of the leaf. This measure will change as you rotate around the entire leaf. The distance rate of change is represented as a value in the frequency domain – a larger value for the spectral peak correlates to a higher rate of change.

Sample Set Output



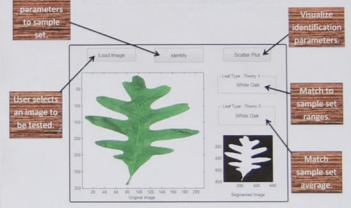
3D Scatter of Sample Leaves with Test Image



Sampling Process



Leaf Identification GUI



Species	Min	Max	Min	Max	Min	Max
Ginkgo	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
Tulip Tree	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
Sweet Gum	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
Sugar Maple	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
Fig Tree	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
White Birch	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
White Oak	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
Sassafras Tree	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
Shagbark Hickory	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
Flowering Dogwood	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000

Results: With this project we were able to lay the essential ground work for an accurate leaf identification system. Building from this foundation, the system could be further developed to identify leaves for a mobile application.

The test image was categorized using two different theories. Theory 1 uses the minimum and maximums, determined from the sample set of images, to establish a range. After determining the different features, those values are then sorted within those ranges.

Theory 2 uses the same ranges as Theory 1 to determine the mean. From this mean, the smallest distance to the test image is determined. The closest one is the resulting category.

Even with two classification processes, we were only able to identify leaves with mediocre success. We found four features that gather a lot of information from the leaves, but we believe one more unique feature is needed to simulate the human perception. We realized there is not one "golden parameter" that can adequately categorize each leaf.

Despite using different theories to narrow down the selection, ties often occurred between each category of leaf. Creating an effective and robust tie-breaker would increase our accuracy.

We also utilized a GUI for easy interaction with the leaf identification system. By simply loading in the leaf and clicking 'Identify', the user can know what the program determines.

Overall Results

