

Small Object Detection in an Outdoor, Time Varying, Dynamic Environment



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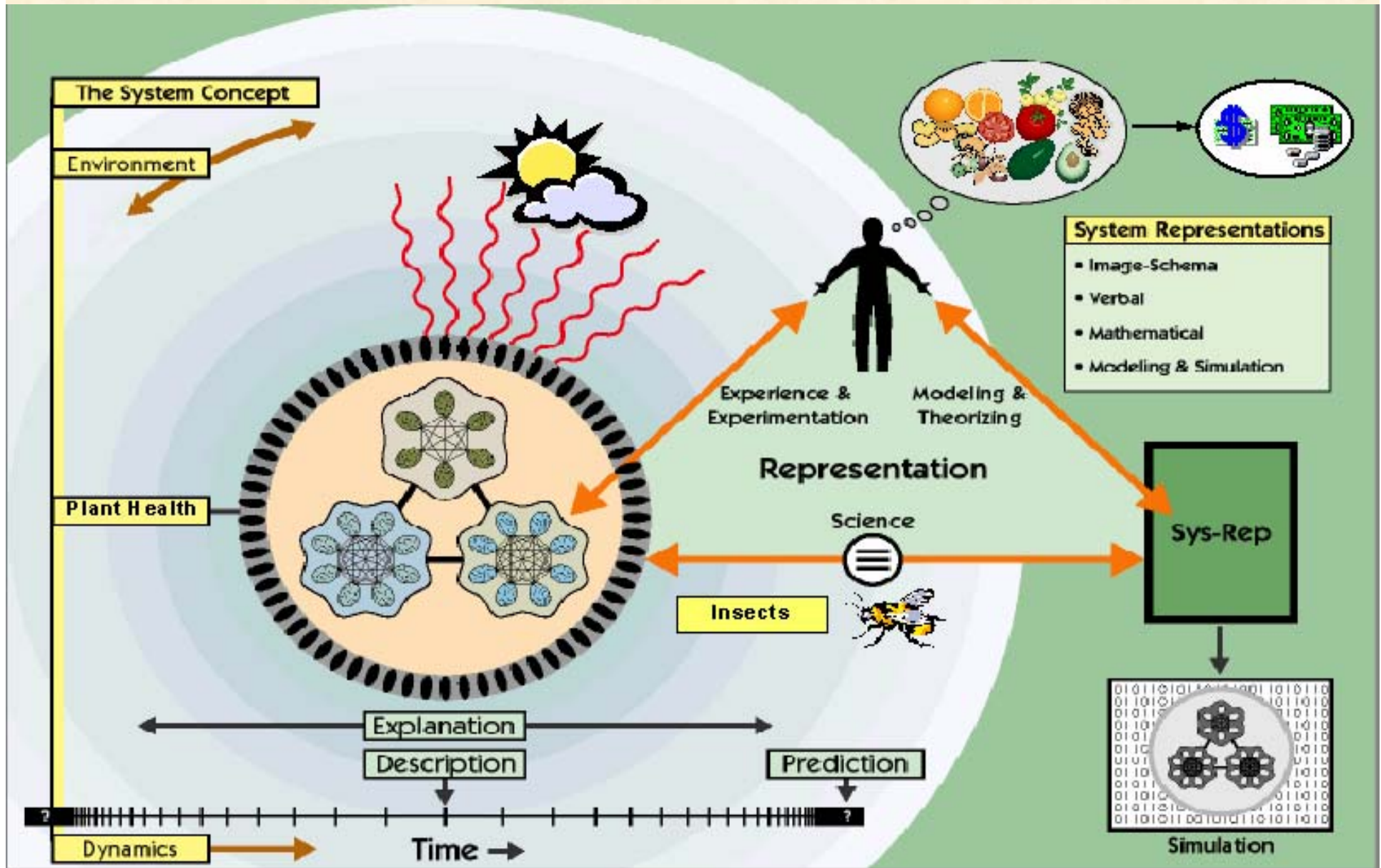
Overview of Presentation

- Introduction
- Research Techniques
 - Results and Findings
- Future Directions
- Summary

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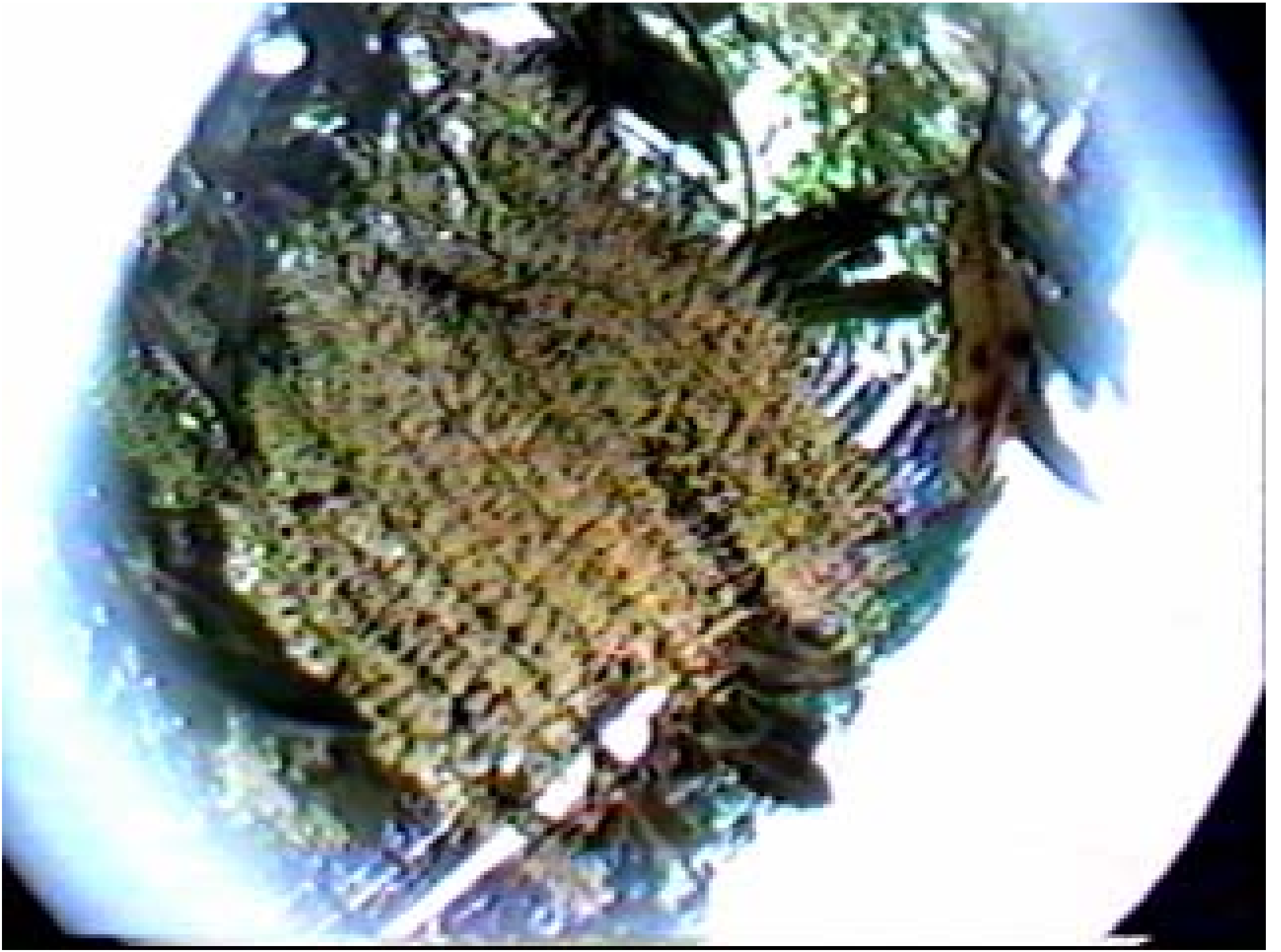
Complex Systems



Research Problem

- To identify small objects (bees) in a video sequence in an outdoor environment
- Faced with three challenges
 - Size of the honeybee
 - Highly dynamic background
 - 3D Environment





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Research Site

- The Macadamia Tree
 - Large evergreen tree which grows in tropical climates
 - Produces the macadamia nut
 - Australia is the world's largest producer,
 - 34,400 tonne in 2001
- Data Collection
 - Portable, CCD cameras placed in a macadamia orchard
 - Video digitized from VCR recordings
 - 320 x 240 pixel resolution @ 24 frames per second



Motivation for Research

- Little research has focused on small object detection rather the majority has focused on tracking people and vehicles
- Existing data gathering techniques were of minimal use when attempting to gather information for use in complex system modeling
- Research into an automated data collection tool for the purposes of complex system modeling

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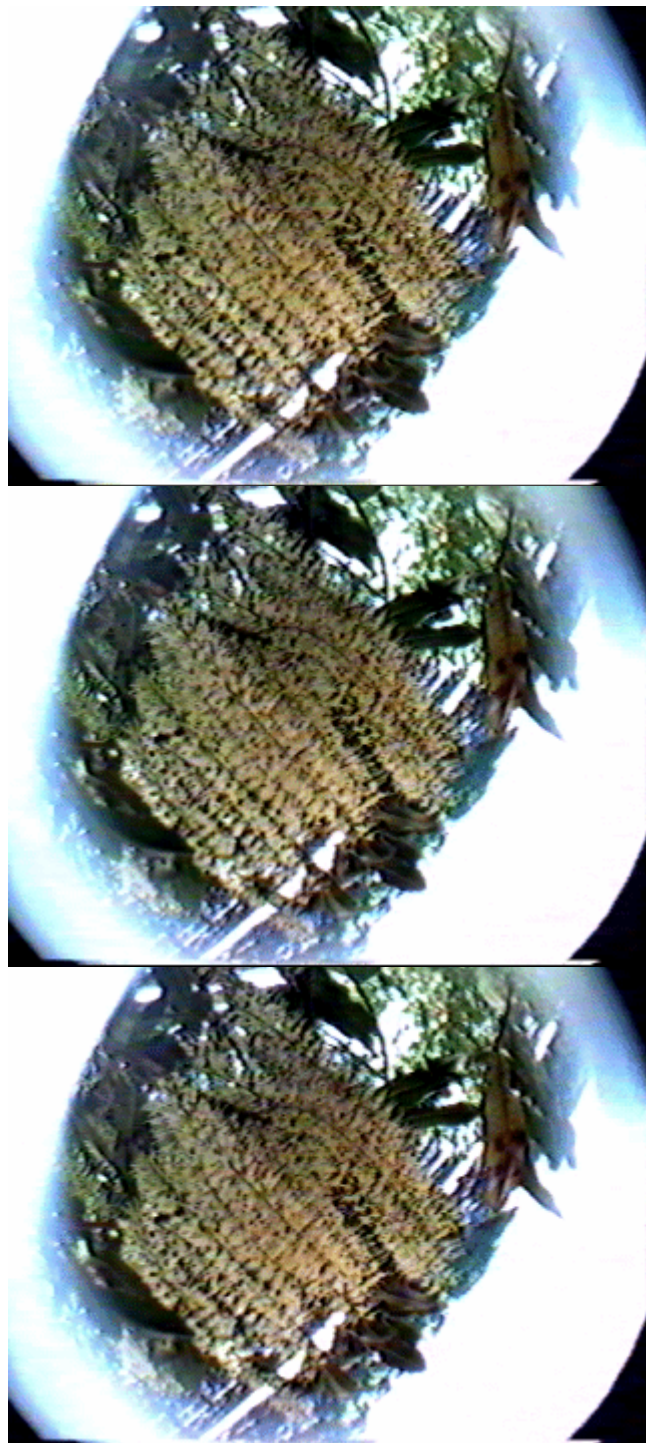
Techniques

- Balch *et al* (2001) used Colour Segmentation
- Colour is not a viable option in outdoors environments (Horn, 1986)
- Edges, shapes, size or texture segmentation is not a viable option



Frame Differencing Techniques in Brief

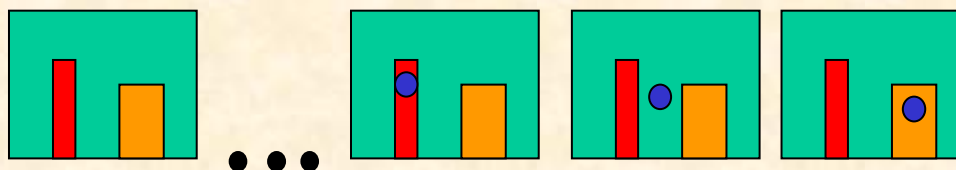
- An adaptive background model generated from $0, \dots, n-1$ images
- Absolute Difference Subtraction
- Noise Filter
- Local Threshold
- Boundary Restoration



Frame Differencing Techniques in Brief

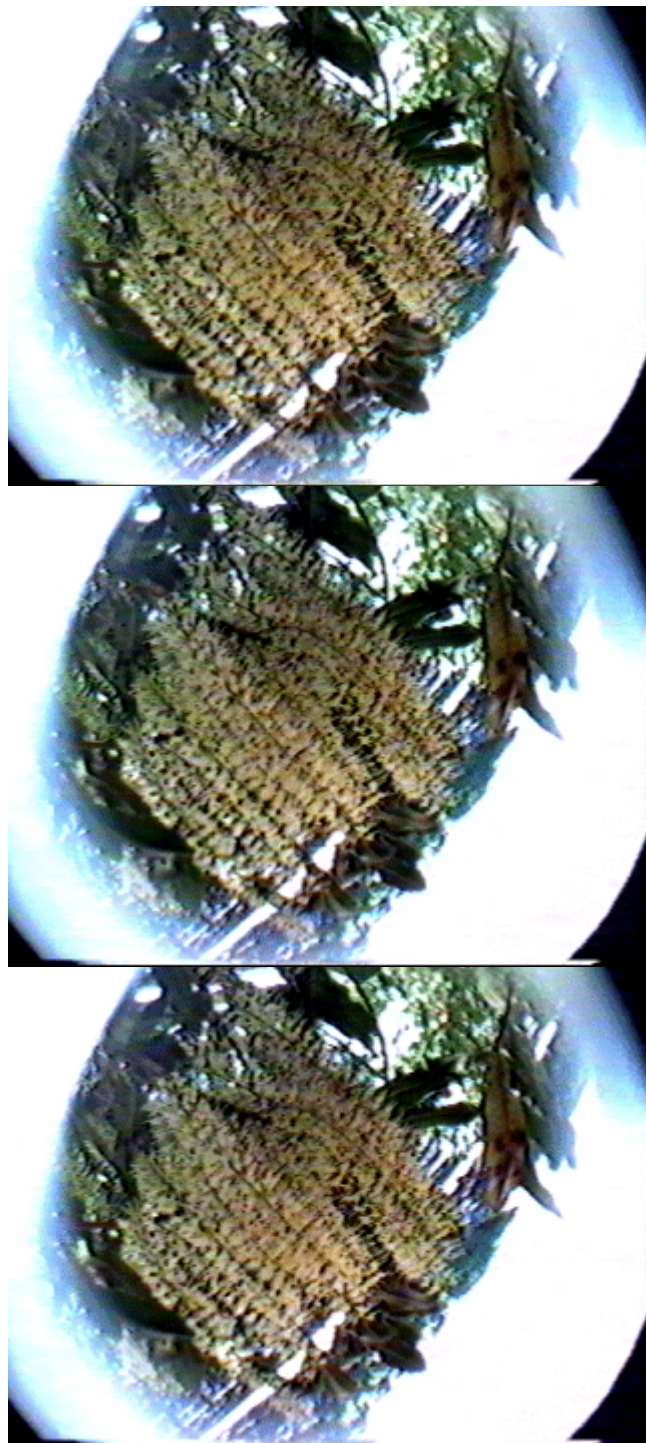
- An adaptive background model generated from $0, \dots, n-1$ images

Consider the following image sequence



Then the background model, B_T is created by:

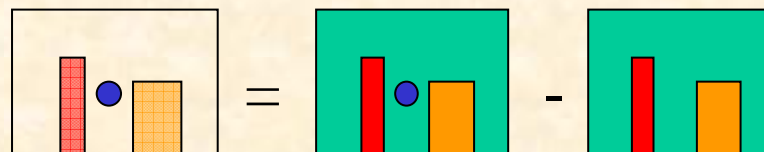
$$B_T = (1 - \alpha) \cdot B_{T-1} + \alpha \cdot I_T$$



Frame Differencing Techniques in Brief

- **Absolute Difference Subtraction**

$$\text{Diff}_T = |I_T - B_{T-1}|$$



- A significant amount of small difference present in the resulting images
- In the images on the left, the lighter the colour, the smaller the difference.

Frame Differencing Techniques in Brief

- **Noise Filter**

$$Th_x = Global_{MED} + 3 \times 1.4826 \times Global_{MAD}$$

where $Global_{MED}$ is the median global pixel value, and $Global_{MAD}$ is the median standard deviation

- A global approach to remove the small differences present in the images
- This technique did not erode the boundary or shape of any detected regions of difference
- However, very small high intensity noise remained

Frame Differencing Techniques in Brief

- **Local Threshold**

$$Th_x = Global_{MED} + NI$$

where $NI = [Global_{MAD} / (Local_{MED} / Global_{MED})]$
and $Local_{Med}$ is a local median pixel value in a 3x3 grid

- A local approach to remove very small high intensity noise by adjusting the global threshold to account for local conditions
- Successfully removed isolated high intensity noise
- However, this technique eroded the boundary/shape of any regions of detected difference

Frame Differencing Techniques in Brief

- **Boundary Restoration**
 - A depth-first recursive algorithm which restores the shape of any detected regions of difference present after the local threshold is applied
 - This is achieved by using the results from the global threshold and applying it to the results after the local threshold
 - Allows the benefits of both the global and local approaches to be used without their associated problems

Frame Differencing

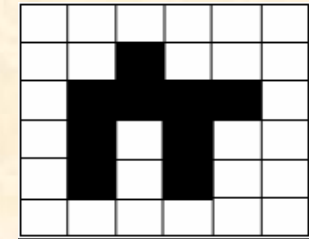
Results & Findings

- A very efficient technique for identifying the location of the bee
- Unable to address problems of rapid illumination change and large sustained movement of branches and flowers

Object Creation

Techniques in Brief

- A depth-first recursive search and merge algorithm is used to create objects

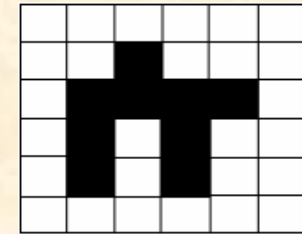


- Then, spatially close objects are merged
- Finally, objects containing less than 2 pixels are removed

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Object Creation

Results & Findings

- By creating objects, the bees are easier to visually identify
- The minimum box was useful for estimating the size of the bee
- The movement of the flowers are detected as a series of very small objects





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Temporal Analysis

Techniques in Brief

- Two background models
 - Forwards constructed from I_{T-11}, \dots, I_{T-1}
 - Backwards constructed from I_{T+11}, \dots, I_{T+1}
- The resulting objects from both models are compared and objects which do not match are removed

Temporal Analysis

Results & Findings

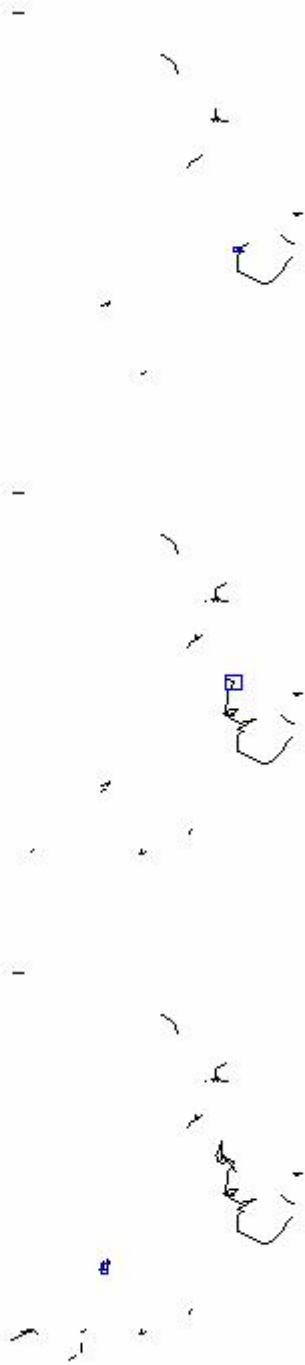
- Very efficient for removing the differences caused by the movement of the flowers
- Excellent results at retaining and extracting the difference caused by the movement of the bee





Object Association Techniques in Brief

- Association between objects with in a maximum distance
- Euclidean Distance between all points in this distance is calculated
- The closest points in Euclidean Distance are joined by an edge and considered the same object in two separate frames



Object Association

Results & Findings

- Good at tracking a bee's path assuming the prior conditions are met
- Also tracks the path of moving objects such as branches, which are difficult to distinguish from bees
- Can create paths from random or isolated noise





More Results



More Results



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Future Directions

- The collection of more video data to allow for further investigation and analysis of the techniques used during this research
- Conditions, such as rain, strong wind and rapid illumination changes are commonly found in outdoor environments and will need to be studied before the development of a fully automated data collection system is practical.

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Summary

- Although further research is required to address a broad range of environmental conditions, the results obtained so far, have proven that small object detection is a feasible technique for addressing the current problems of data collection.
- **Published Work Arising from this Research**
Estivill-Castro, V. Lattin, D. Suraweera, F. & Vithanage, V (2003) *Tracking Bees- A 3D, Outdoor Small Object Environment*, To be published in the Proceedings of the International Conference on Image Processing, Barcelona, Spain, September 14th – 17th

Questions?