

# Automatic Insect Detection and Classification on Sticky Traps

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### Introduction

Timely pest control decision is critical for the success of the nursery program to meet seed target. This process is time consuming and could result in a delayed decision of control. Automating the process of pests estimation enables faster and better data driven decision making, and ensure the success of nursery pest management.

## Solutions Evaluation

Several approaches for automatic decision making process by analyzing the sticky traps have been proposed and tested. A summary of those solutions and their weakness is provided here:

- Estimate the covered area of the sticky traps and decide based on it's condition. If sticky is too dirty then there are many insects on the traps and should be sprayed.
  - × The plant residual, dirt and dust mislead the decision process.
  - × Types of pests and frequency of each types are important for decision making process.
- Segment insects and define types of insects based on their predefined size and then count them and decide.
  - × Same insect species can have different size.
  - × Different insect species can have same size.
- Segment insects with image processing methods and classify them with feature extraction and machine learning techniques.
  - Traditional feature extraction can not extract rich features to detect the subtle differences between insect species.
- Segment insects with image processing methods and use a Convolutional Neural Network with Transfer Learning to classify them.
  - × Due to inefficient region proposal process this solution is very slow.

After an exhaustive search around various solutions we adopted a Faster Region-based Convolutional Neural Networks for the problem.

## Research Objective

Design and build a model that automatically detects the pests on yellow sticky card, and report the number of each species. Associate the counting with an unique identifier assigned to it.





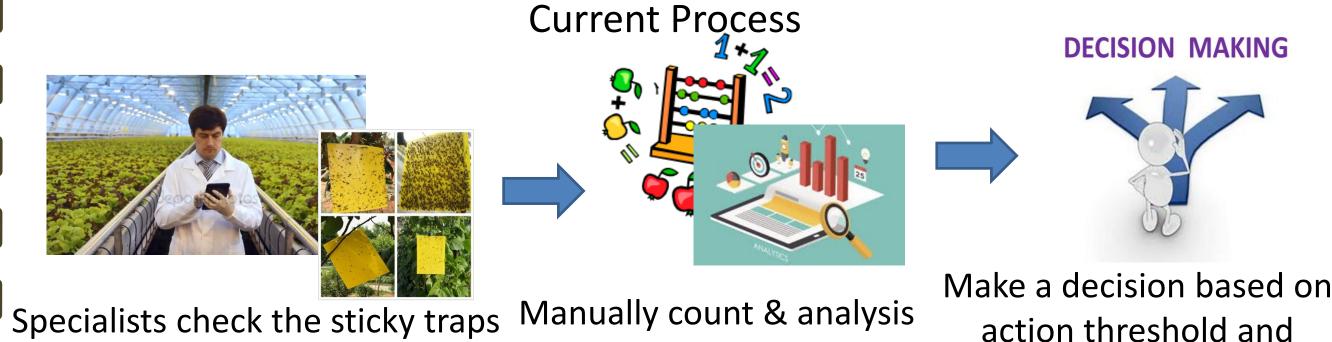








## Process Engineering

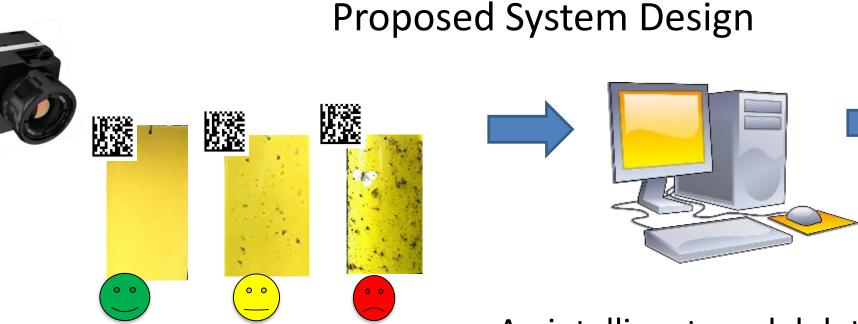


periodically

the insects

action threshold and domain knowledge

Spray or not



An automated camera scans the sticky traps periodically

An intelligent model detects and counts the insects in the sticky traps Make a decision based on action threshold and domain knowledge

#### System Schema

Scan the sicky traps & barcode

**Preprocessing** extracting the yellow part & barcode

Insects type detection & counting

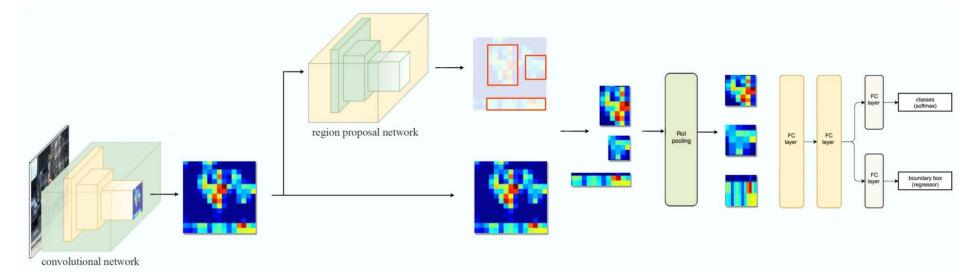
Intelligent decision making to spray

Store the result & barcode in database

## Research Contribution

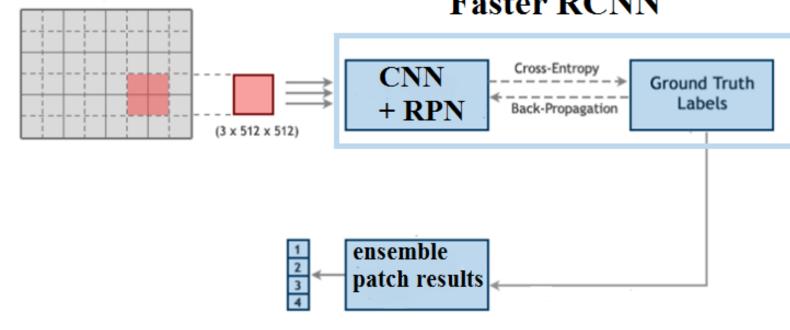
The proposed model for pests detection is based on Faster RCNN model on the backbone of inception V2 which is pretrained on COCO dataset. It is implemented with TensorFlow's Object Detection API.

#### Faster RCNN Architecture



The model breaks down the whole sticky image to multiple overlapping patches and applies Faster RCNN on each patch separately to address the difficulty of the small object detection. Then the outcome of all patches combines together to predict the type and population of insects in the sticky trap.

#### Proposed Architecture **Faster RCNN**



The system prediction on some test data is shown below.

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	IMG_9269	2	61	5	27		thrips: 9	thrips: 99%	100	fungus: 99%	9%	1
	IMG_9268	18	2	20	6						/	
	IMG_9272	244	14	238	8							
	IMG_9274	3	29	6	27				4			
	IMG_9275	21	6	24	13				*			
	IMG_9276	21	7	15	2			-1			fungus gnats	Thri ps
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