## POL-ID: AI-Based Honey Authentication

### Introduction

Honey authentication is critical for consumer trust and trade. Manual identification of pollen under a microscope is time-consuming, labour-intensive and error-prone. Our goal is to automate pollen detection and classification to support apiculture and food traceability.

### **Objectives**

We aim to develop a region-specific deep learning pipeline capable of detecting pollen grains in microscopy images, classifying South African pollen taxa, and clustering novel or unlabeled types. This integrated approach will be tailored to South African pollen.

### Methodology

We use **YOLOv11** to detect pollen grains and generate bounding boxes in microscopy images. Classification is carried out with three complementary models: a **Convolutional Neural Network (CNN), a Vision Transformer (ViT), and a hybrid CNN–Transformer model**. To identify pollen types not present in training, **HDBSCAN** clustering is applied to feature embeddings, enabling the discovery of novel groups. Performance is evaluated using **mAP**, **precision**, **recall**, **F1-score**, **and silhouette score**, ensuring both accurate classification and meaningful clustering results.

### Detection

D

mAP@

%8.8%

# pollen 1.0 95 0004 pollen 0.9 pol

Results

CNN
Classifier
94.43% test
accuracy
0.916 F1-score

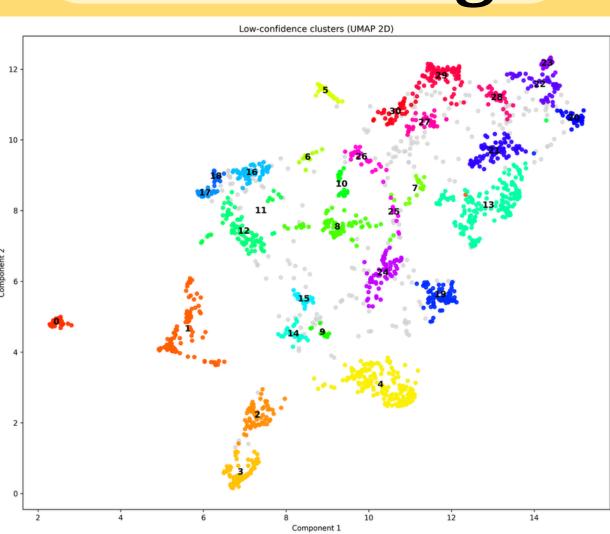
### Classification

Hybrid
Classifier
96.5% test
accuracy
0.960 F1-score

VIT
Classifier
92.64% test
accuracy
0.902 F1-score

92.8% mAP@50-95

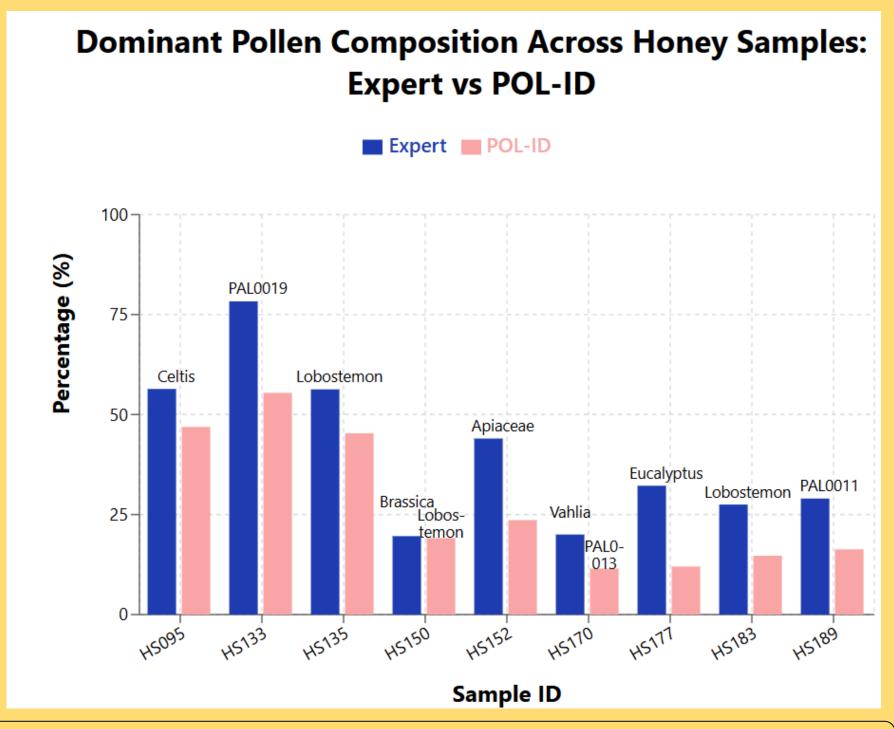
### Clustering



Noise fraction = 12.8%

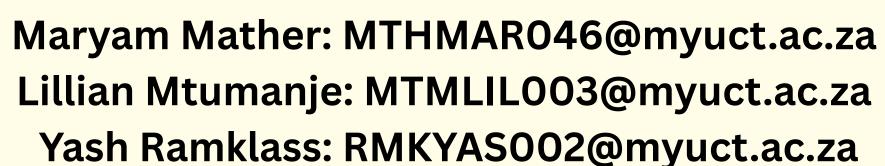
### Pipeline

- Best performing pipeline was the CNN pipeline
- All 9 samples
   correctly
   classified
   according to
   ICBB standards



- Many clusters were dominated by a specific pollen class
- Class imbalances need to be reduced + increased dataset sizes

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