

SKY CROP

in cooperation with:



Pain: Client Needs

What information do you need to start a rice or sugarcane mill in a new region?



Rice and
Sugarcane
Mill Owners



FIRST - Field segmentation data is key

- Which crops are now being grown in individual fields locally?

THEN - Use the field segmentation data to estimate:

- Mill crop input amounts (local rice/sugarcane)
- Amounts of fertilizer/pesticides to sell to rice and sugarcane farmers
- Size of mill to be built, equipment, and required investment

Pain: Current Hurdles

To identify crops in individual fields, mill owners use:

- Existing government land and crop databases
 - ✗ Not up-to-date
- Creating own new database
 - ✗ Take a lot of human resources, time, and money
 - ✗ Visiting fields on foot can be costly and unwelcome in a state of COVID

Solution: SKY CROP



Our SKY CROP app can identify the crop type (rice vs. sugarcane):

01

Remotely (satellite)

02

Using latest data

03

At a lower cost

Solution: Roadmap

01

Segment fields



02

Identify crops that are being grown
in real time

03

Estimate production amounts of local fields

04

Estimate amounts of fertilizer and
pesticides to be sold to farmers

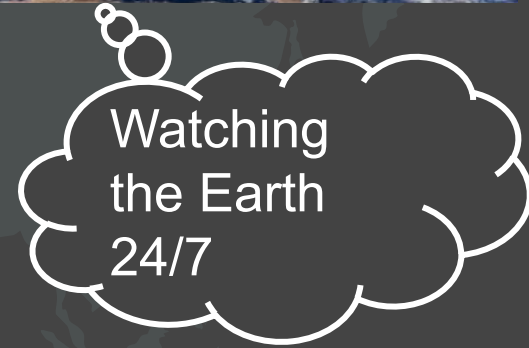
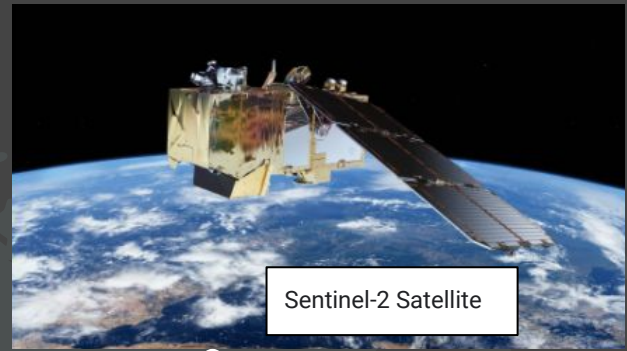
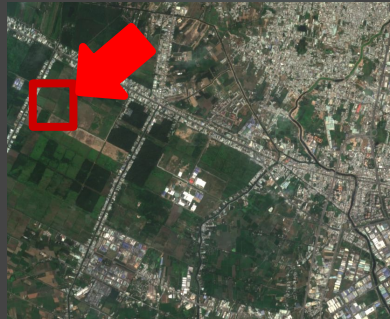
Data collection

Data provided by



- Vietnam - Segmented rice field imagery
- Thailand - Segmented sugarcane field imagery

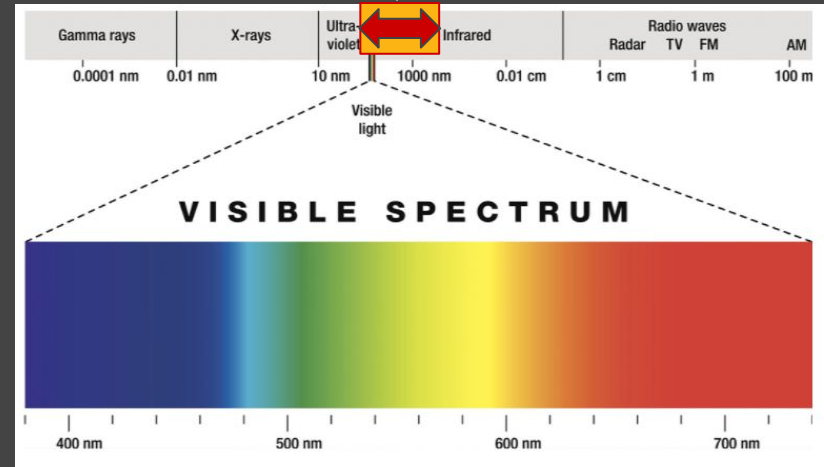
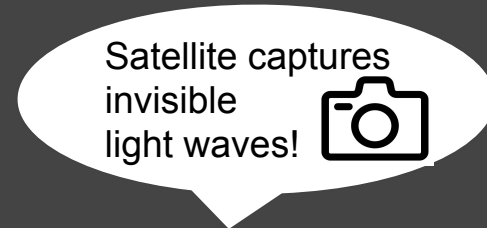
Humans cannot see the difference between rice/sugarcane fields with the **naked-eye**



Satellite Imagery: Band Data

Sentinel-2 Satellite captures 13 spectral bands: visible (red, green, blue), NIR, red edge, SWIR, and atmospheric bands

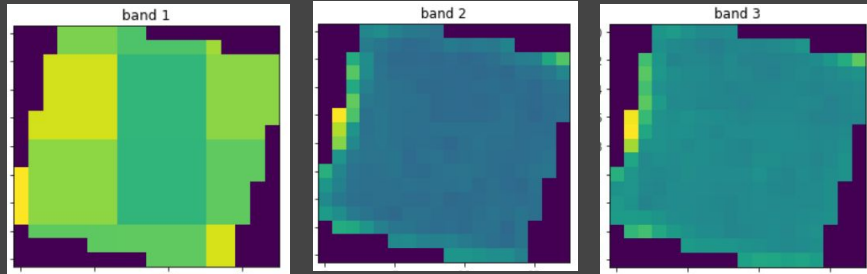
Band data can be used to assess the state and change of vegetation, soil, and water cover over time



Data Sample

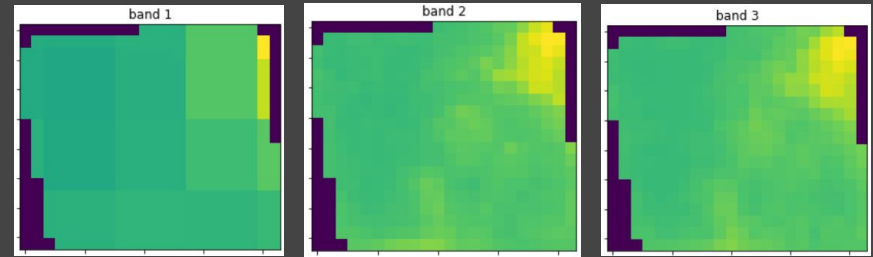


Rice field



Band 1: [0, 0, 819, 819, 0, ...]
Band 2: [0, 0, 735, 824, 0,]
Band 3: [0, 0, 1032, 1138, 0, ...]
□
Band13: [0, 0, 642, 642, 0, ...]

Sugarcane field



Band 1: [684, 577, 577, 577, 577, ...]
Band 2: [751, 496, 454, 401, 388, ...]
Band 3: [1106, 865, 697, 700, 724, ...]
□
Band13: [1586, 1521, 1521, 1401, 1401, ...]

Numeric information of each band was used to train our model

Band Combinations

False Color Index

$(B7 + B6 + B4):$

Useful for:

- (1) Visualizing areas of **dense vegetation**
- (2) Identifying **vegetation types**

Agriculture Index

$(B11 + B8 + B2):$

Useful for:

- (1) Monitoring the **health of crops**
- (2) Highlighting dense vegetation (dark green)

Green Normalized Difference Vegetation Index (GNDVI)

$(B3 - B8) / (B3 + B8):$

Sensitive to the variation of **chlorophyll content** in the crop.

Train Classification Model

START

Baseline model

Logistic
regression model

Add more features
and data, try
different algorithms

Train Model



GOAL

Best model

KNN model

**Accuracy:
95%**

SKY CROP: <http://skycrop.herokuapp.com/>

Challenges

- Complex (unstructured) data
 - Satellite imagery data extraction, manipulation, and preprocessing
- Selecting important features
 - Which factors impact model scores?
 - Which feature is important?
- Selecting the best performing model
 - KNN, random forest, logistic regression, SVM, neural network, SGD
- Creating a simple user interface and easily input field data

What's Next?

- More classes (new crops - cassava, corn, wheat, etc.)
- New geographies (Southeast Asia, Indonesia, India)
- Estimate production yield
- Automate segmentation of individual fields
- Create simple, intuitive user interface for mill owners



Our Team



Yuki



Hong



David

In cooperation with



An aerial photograph of a vast, green agricultural landscape. The foreground is dominated by a field of young corn plants, showing distinct rows. A dirt road or path runs diagonally across the middle of the frame. In the background, there's a line of trees and more distant fields under a bright, cloudy sky. The overall scene is peaceful and rural.

Thank you!