



Dr. Terry Bates

Cornell AgriTech
New York State Agricultural Experiment Station

EFFICIENT



VINEYARD

The goal of Efficient Vineyard Project is to research, develop, and implement digital agriculture solutions for commercial U. S. grape production.

The multi-institutional and multi-disciplinary project was supported by the USDA-NIFA-Specialty Crop Research Initiative program from 2015-2020 and led by Dr. Terry Bates of the Cornell Lake Erie Research and Extension Laboratory.

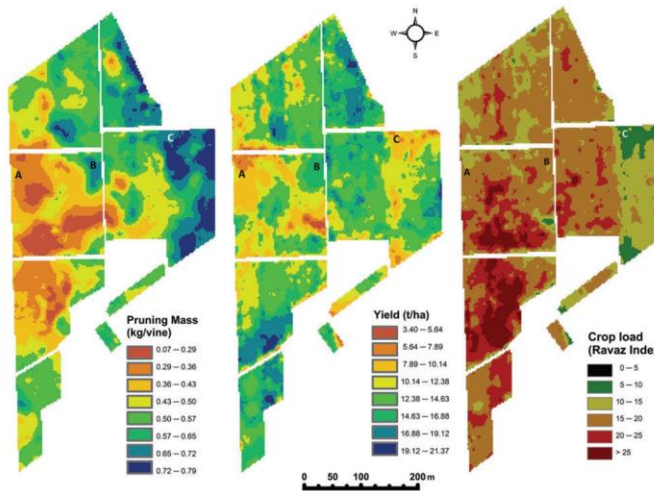
The Efficient Vineyard theme will continue under a new SCRI project on precision nutrient management led by Washington State University.

The Measure, Model, Manage Theme

Spatial Data Driven, Variable-Rate, Mechanized Vineyard Crop Load Management



Measure
Sense
Data



Model
Plan
Information



Manage
Act
Decisions

Establish your farm and goals

Management Outcomes need to drive Precision Viticulture

Locally determined, one size does not fit all

Lake Erie Concord Production

Lower Profit Margins

Ownrooted – Not Irrigated

Becoming more Mechanized

Focus on Crop Load Management
(Designed overcropping with VR vine balance)



Lake Erie Riesling Production

Higher Profit Margins

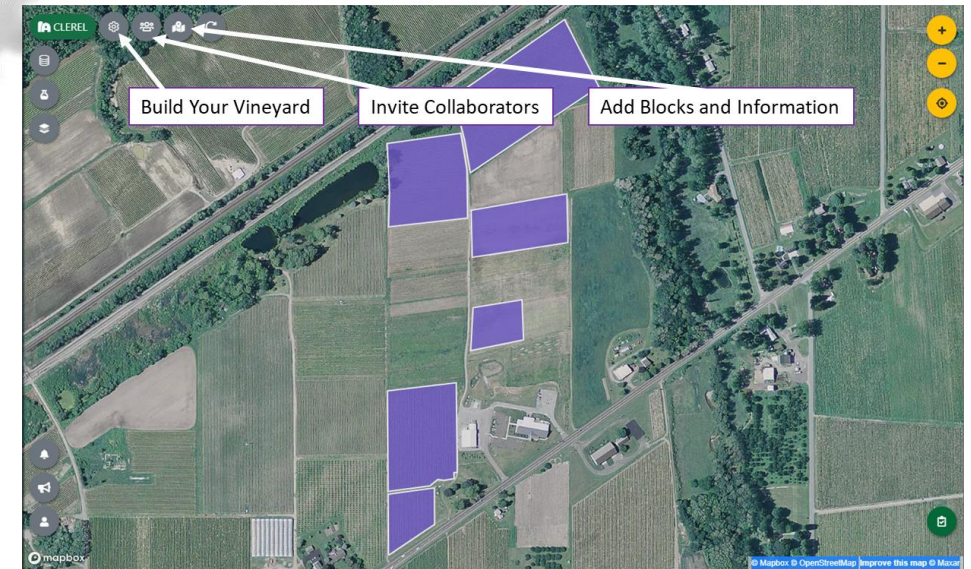
Grafted on more vigorous rootstocks

Predominantly manual management

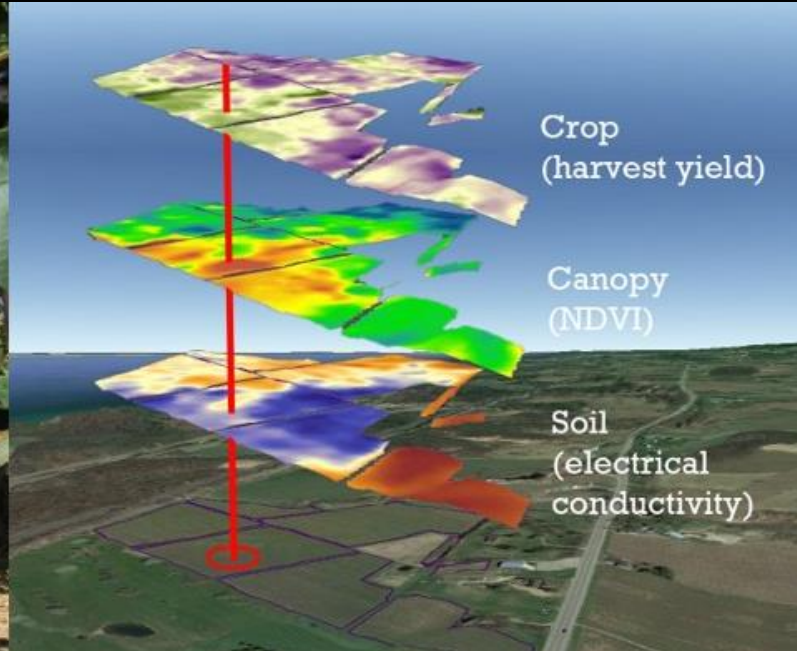
Focus on Canopy Management
(Canopy density, cluster exposure)



Build your farm in MyEV and
import the data you want



The Efficient Vineyard Approach



Measure vineyard soil, canopy, and crop characteristics using mobile field sensors

Model multi-layer spatial data needed for perennial cropping systems

Manage vineyards by integrating spatial information with variable-rate technology

Sensors for collecting vineyard spatial information



Soil DualEM



Canopy CropCircle



Crop CMU, Bloomfield



Harvest Yield/Brix OXBO



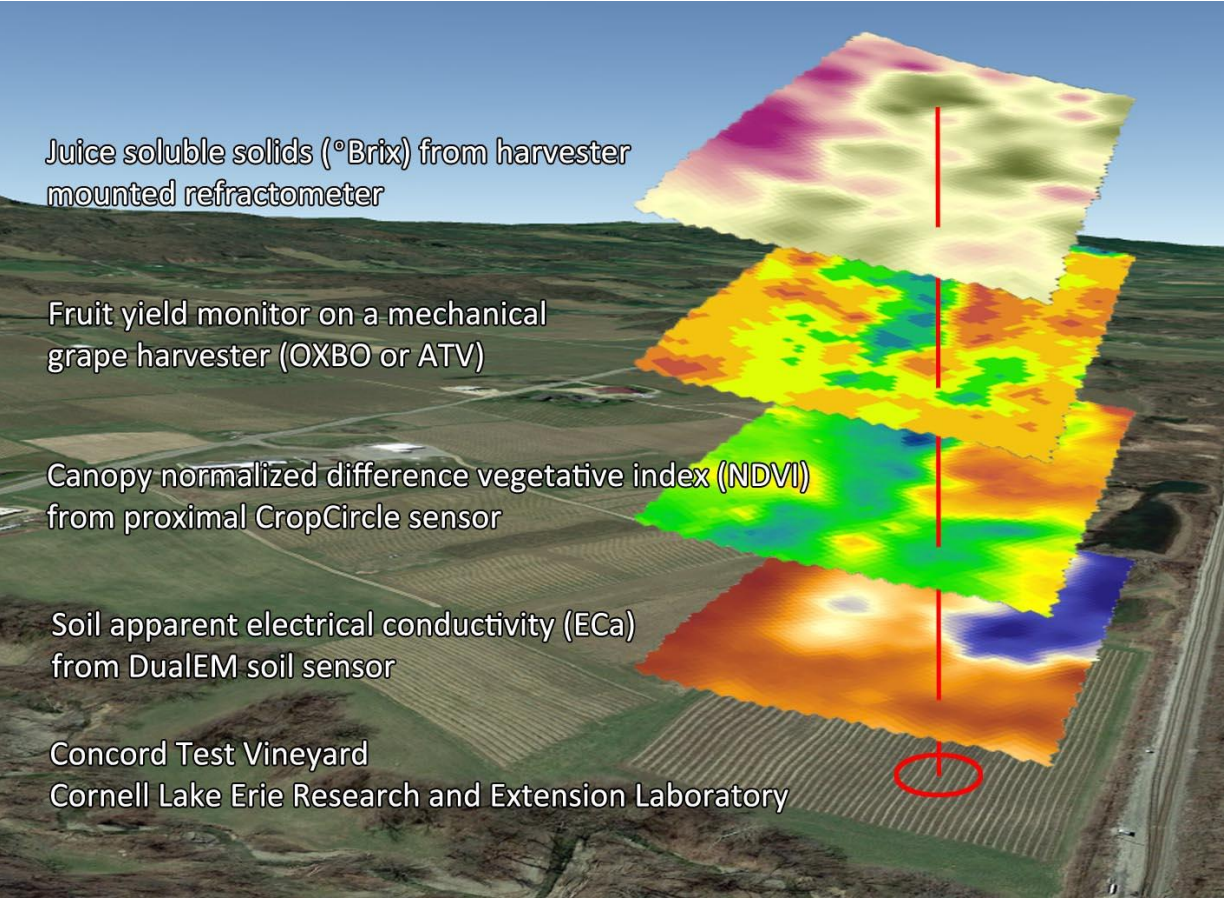
Vine Nutrients RIT



Fruit Metrics Felix Instruments



Anything You + MyEV Data Collector



Import Spatial Data

Sensing Tools in the Toolbox

Collecting relevant spatial data...translating to viticulture information

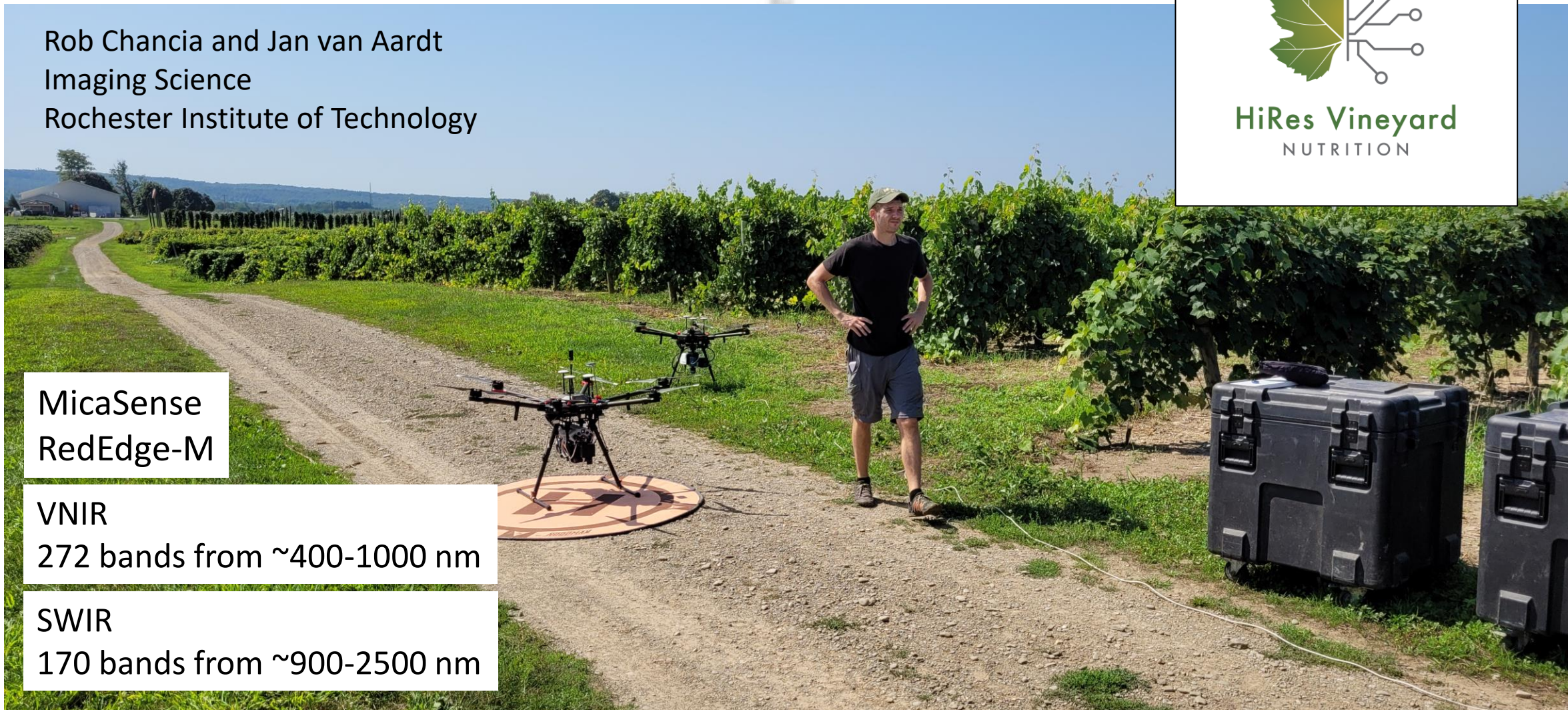
Rob Chancia and Jan van Aardt
Imaging Science
Rochester Institute of Technology



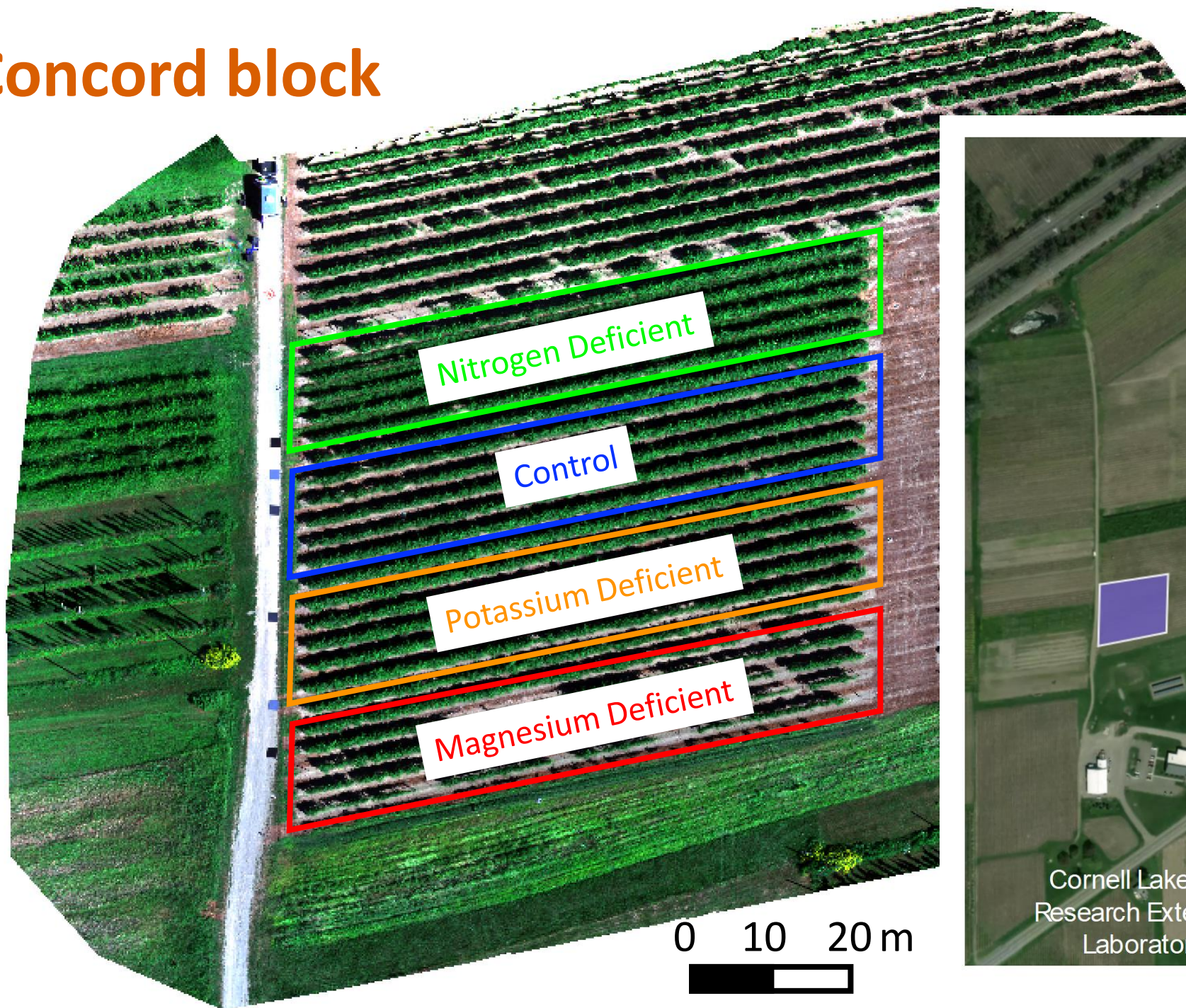
MicaSense
RedEdge-M

VNIR
272 bands from ~400-1000 nm

SWIR
170 bands from ~900-2500 nm



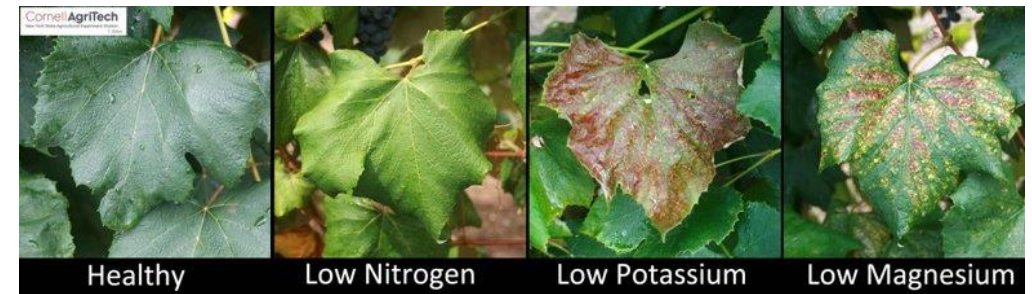
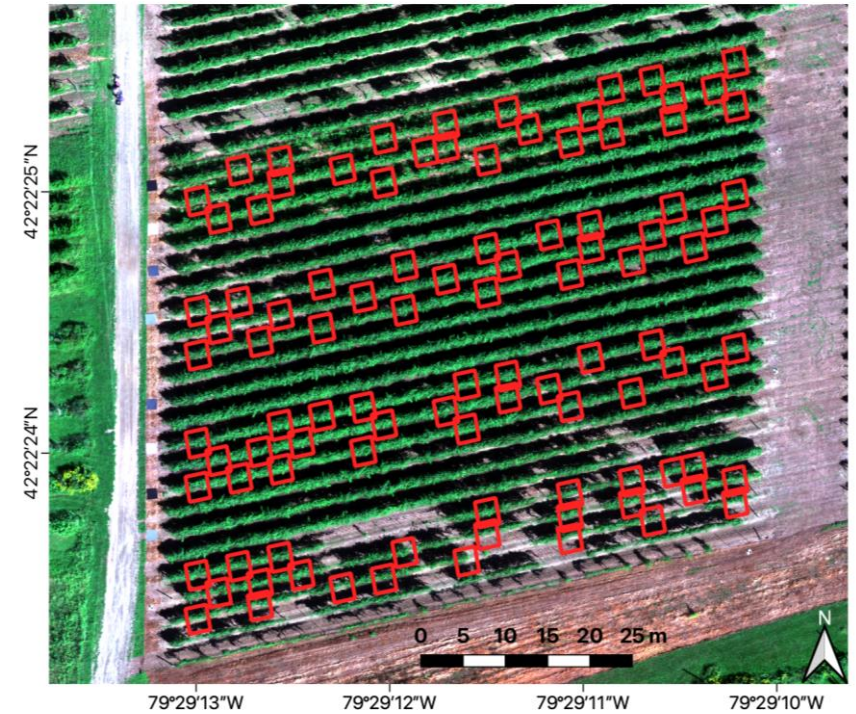
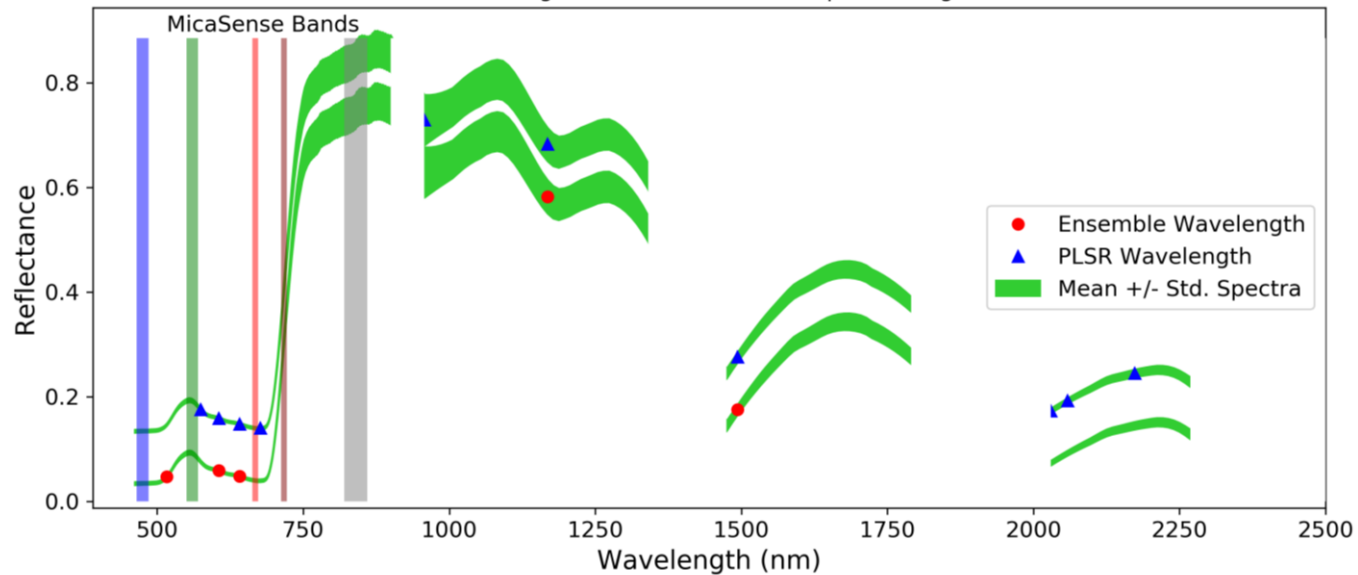
1-acre Concord block



Leaf Nitrogen (%) wavelength selection

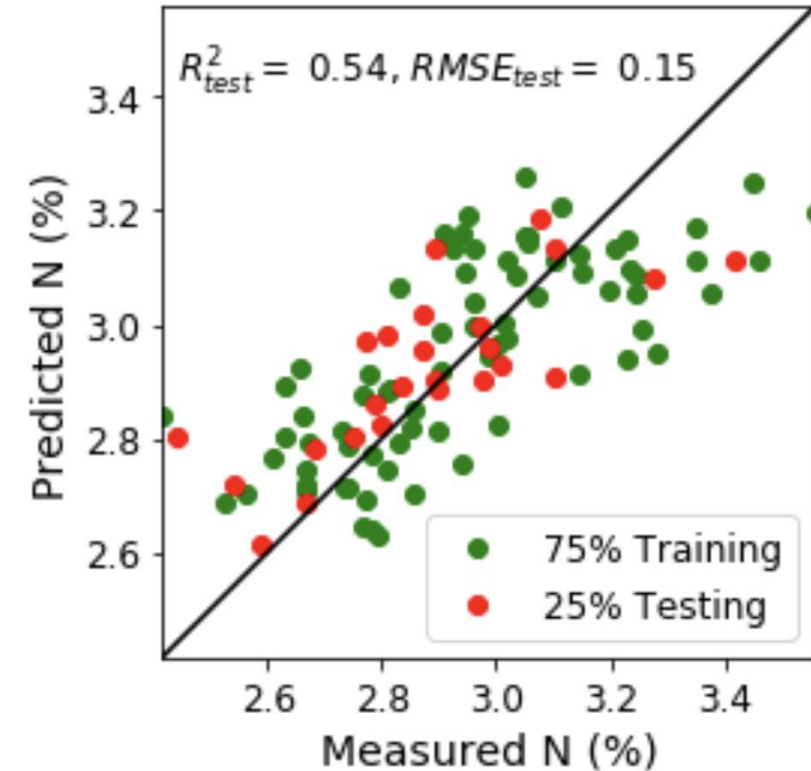


Leaf Nitrogen - Ensemble & PLSR top wavelengths



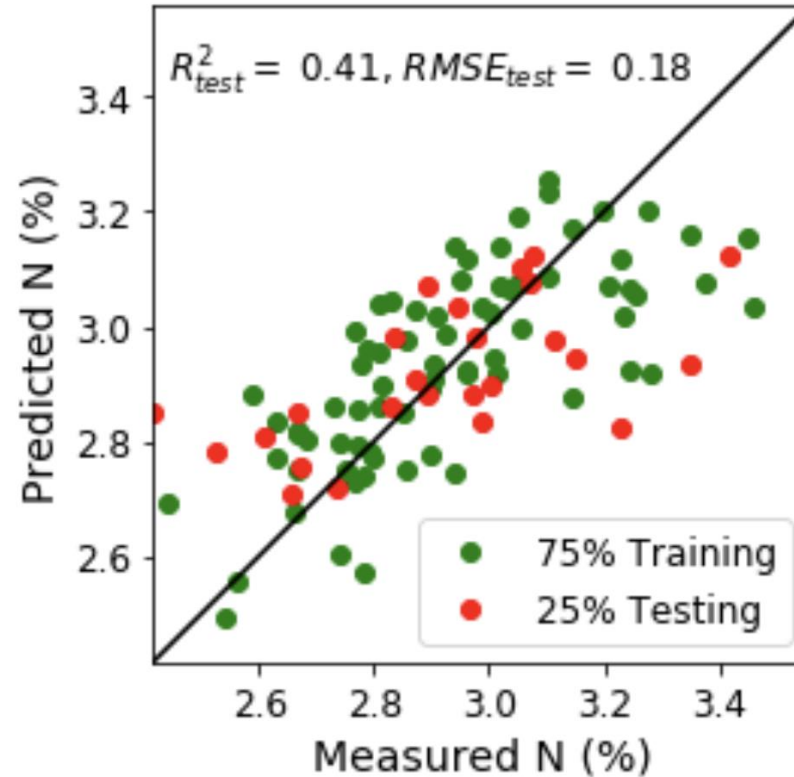
Leaf Nitrogen (%) regression results

VNIR+SWIR: 5 wavelength ensemble selection

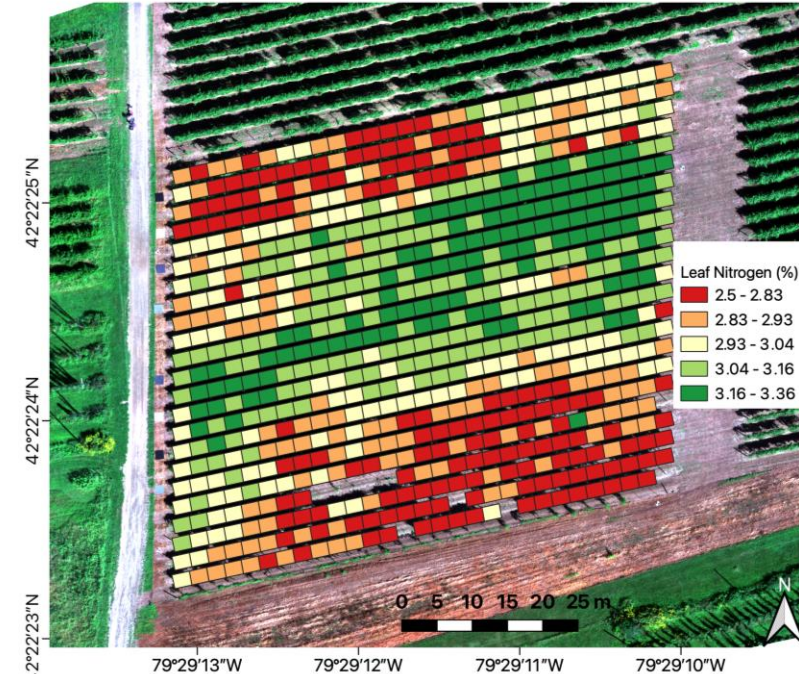


LOOCV: $R^2 = 0.44$, $RMSE = 0.17$

MicaSense



LOOCV: $R^2 = 0.41$, $RMSE = 0.18$



Railroad Block Nitrogen

Vine Nitrogen (%)

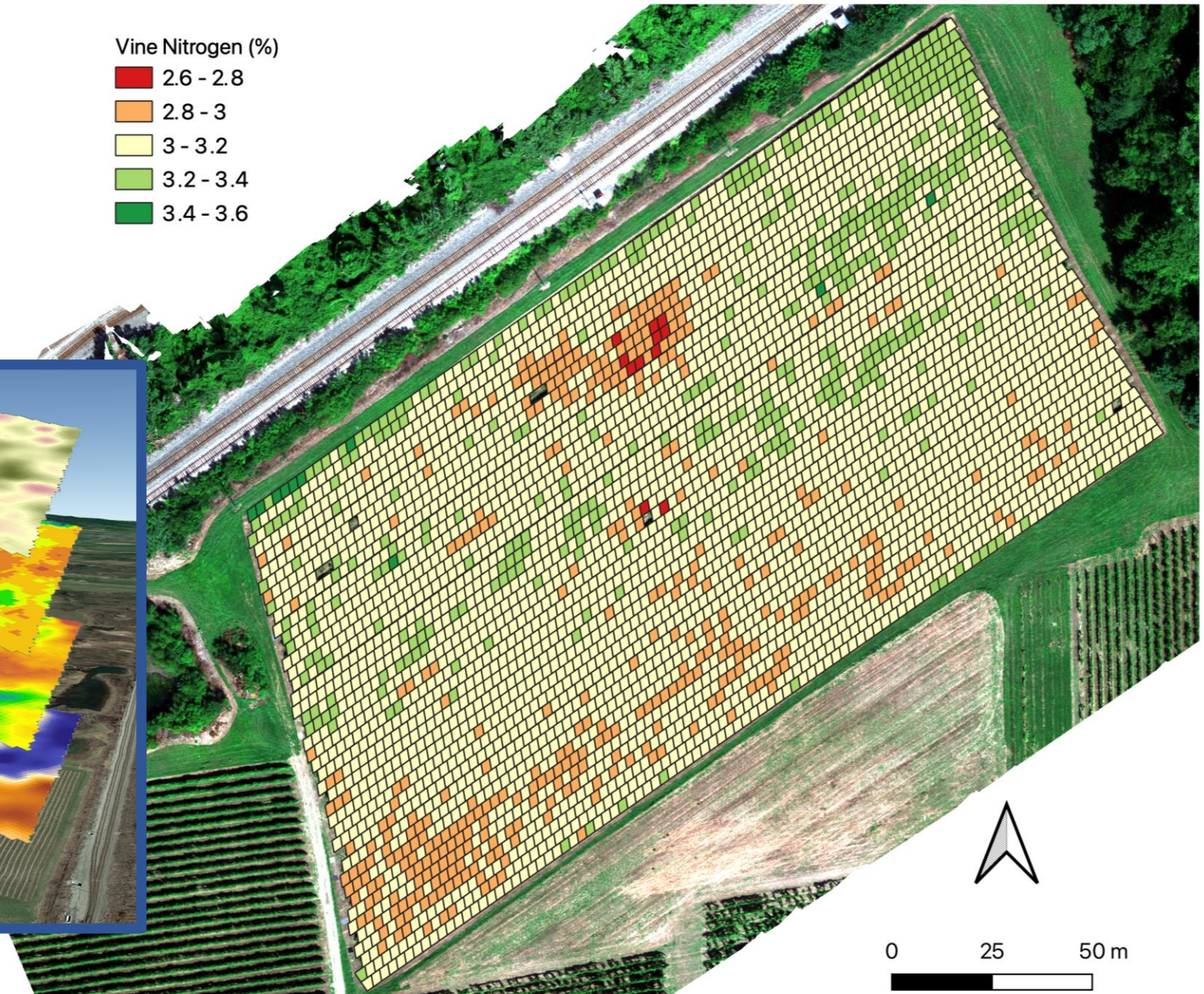
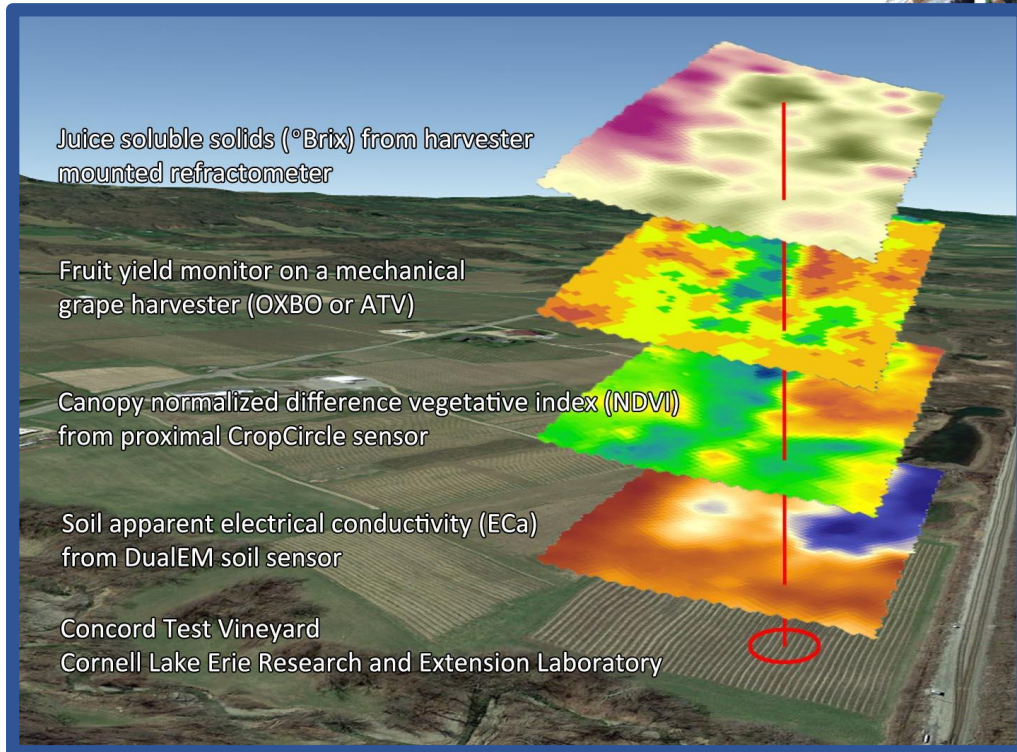
2.6 - 2.8

2.8 - 3

3 - 3.2

3.2 - 3.4

3.4 - 3.6



Import Spatial Data

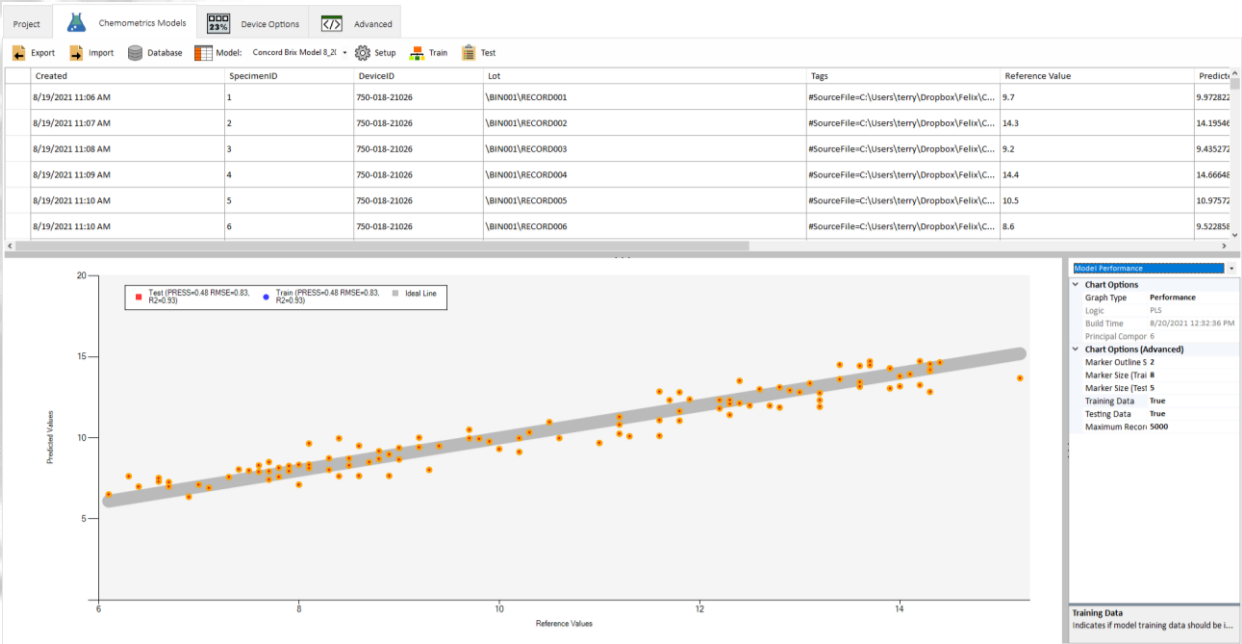
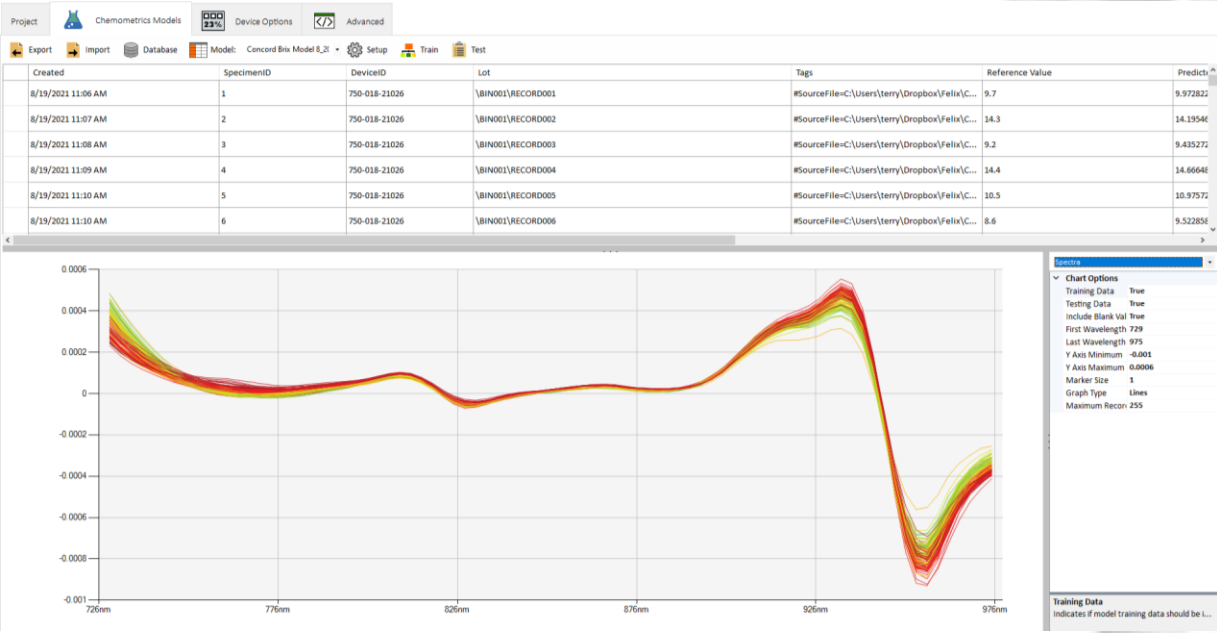
Sensing Tools in the Toolbox

Collecting relevant spatial data...translating to viticulture information



Import Spatial Data

Sensing Tools in the Toolbox



Sensing Tools in the Toolbox

Collecting relevant spatial data...translating to viticulture information

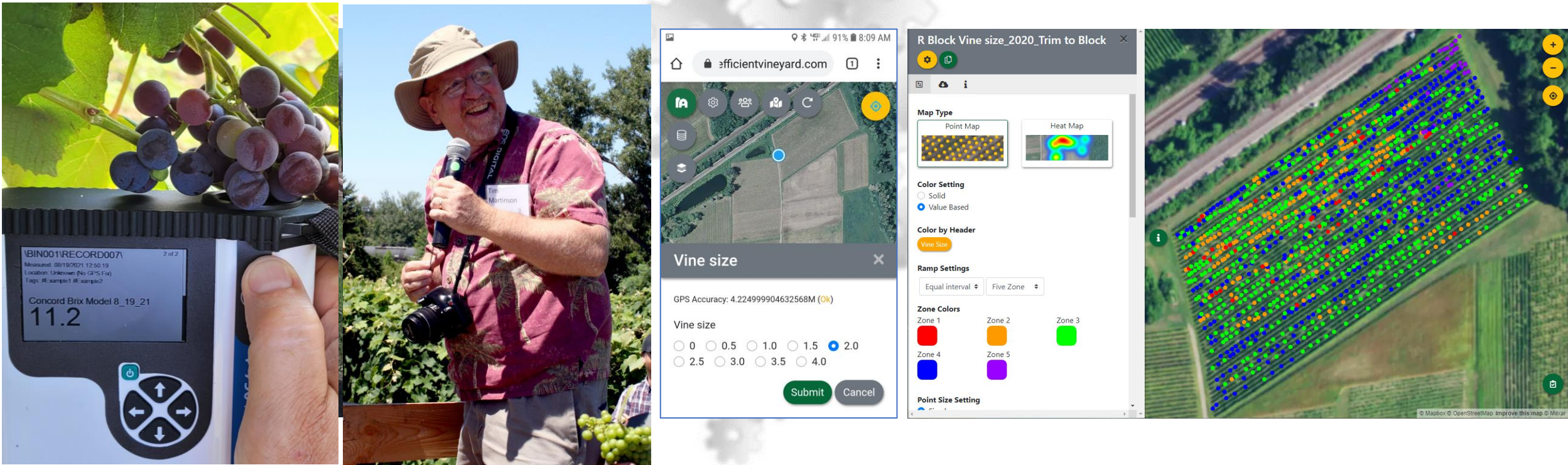


Import Spatial Data

Sensing Tools in the Toolbox

Collecting relevant spatial data...translating to viticulture information

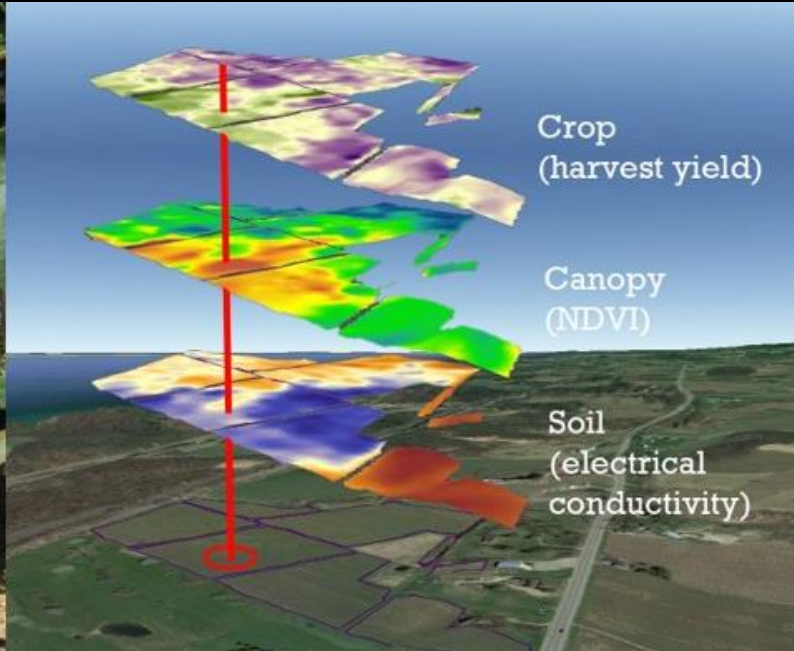
The MyEV Data Collector tool and the “Human” Sensor



The Efficient Vineyard Approach



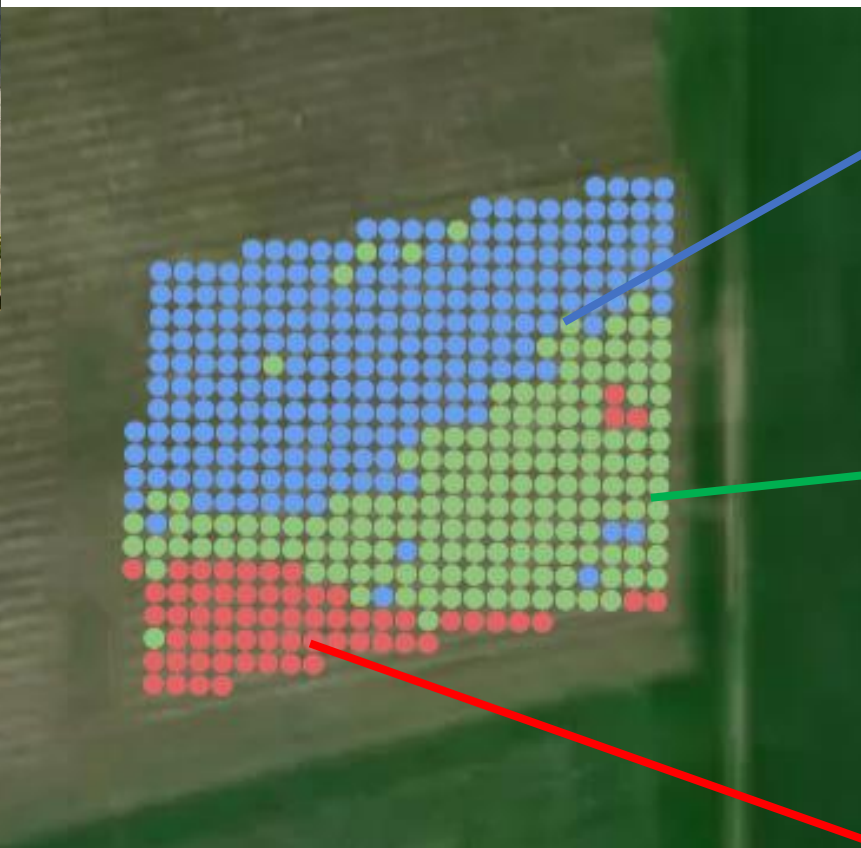
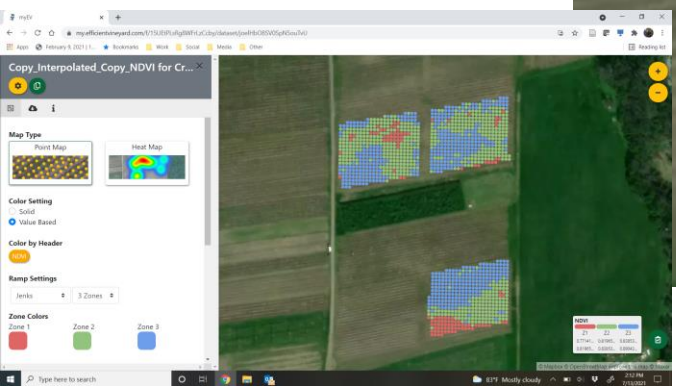
Measure vineyard soil, canopy, and crop characteristics using mobile field sensors



Model multi-layer spatial data needed for perennial cropping systems



Manage vineyards by integrating spatial information with variable-rate technology



3 pound

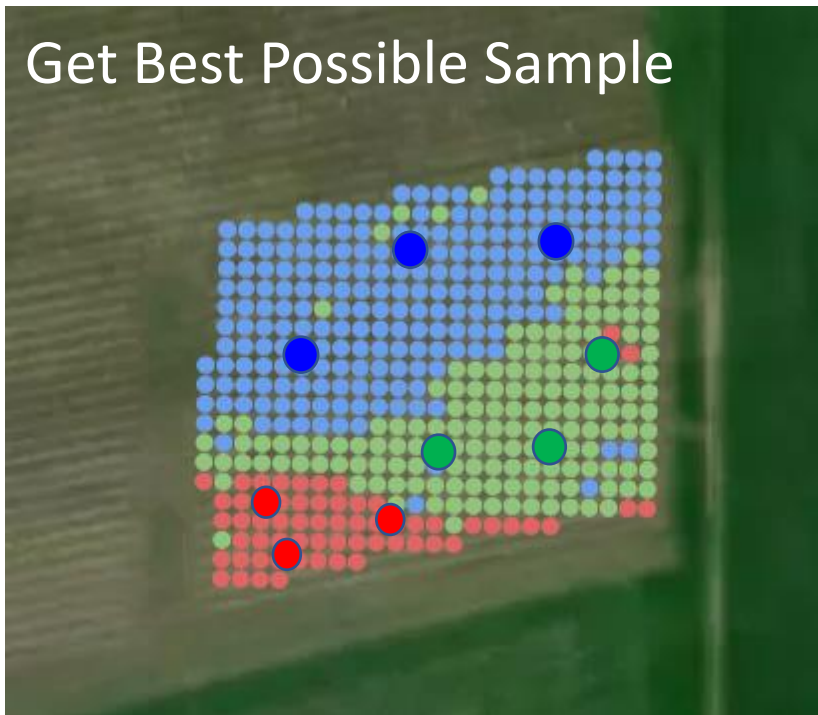


2 pound

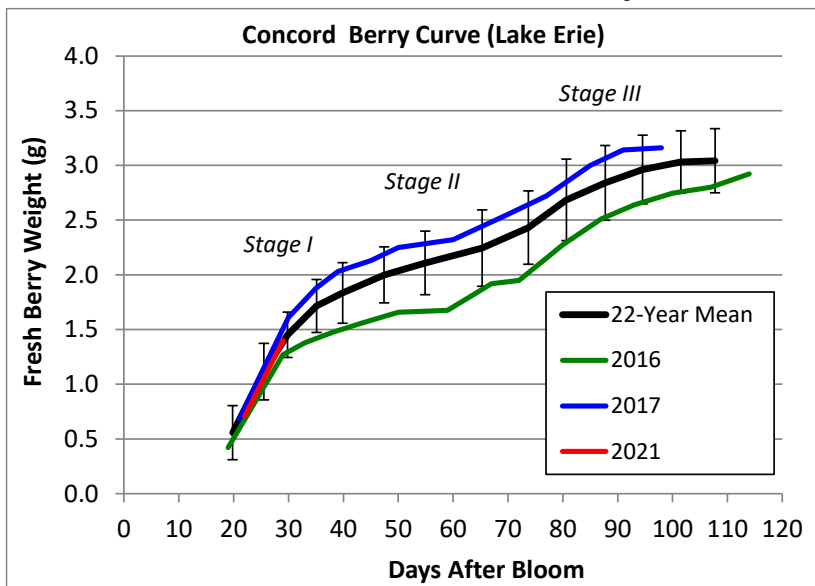


1 pound

Get Best Possible Sample



Use Concord Fresh Berry Curve



Dr. Terry Bates: Crop Estimation and Thinning Table: 7/16/2003

Pounds of Fruit Removed in 1/100th of an Acre	Time of Season														
	20DAB		25DAB		30DAB		40DAB	50DAB	Veraison					Harvest	
	% of Final Berry Weight														
	20	25	30	35	40	45	50	55	60	65	70	75	80	90	100
10	2.5	2.0	1.7	1.4	1.3	1.1	1.0	0.9	0.8	0.8	0.7	0.7	0.6	0.6	0.5
20	5.0	4.0	3.3	2.9	2.5	2.2	2.0	1.8	1.7	1.5	1.4	1.3	1.3	1.1	1.0
30	7.5	6.0	5.0	4.3	3.8	3.3	3.0	2.7	2.5	2.3	2.1	2.0	1.9	1.7	1.5
40	10.0	8.0	6.7	5.7	5.0	4.4	4.0	3.6	3.3	3.1	2.9	2.7	2.5	2.2	2.0
50	12.5	10.0	8.3	7.1	6.3	5.6	5.0	4.5	4.2	3.8	3.6	3.3	3.1	2.8	2.5
60	15.0	12.0	10.0	8.6	7.5	6.7	6.0	5.5	5.0	4.6	4.3	4.0	3.8	3.3	3.0
70	17.5	14.0	11.7	10.0	8.8	7.8	7.0	6.4	5.8	5.4	5.0	4.7	4.4	3.9	3.5
80	20.0	16.0	13.3	11.4	10.0	8.9	8.0	7.3	6.7	6.2	5.7	5.3	5.0	4.4	4.0
90	22.5	18.0	15.0	12.9	11.3	10.0	9.0	8.2	7.5	6.9	6.4	6.0	5.6	5.0	4.5
100	25.0	20.0	16.7	14.3	12.5	11.1	10.0	9.1	8.3	7.7	7.1	6.7	6.3	5.6	5.0
110	27.5	22.0	18.3	15.7	13.8	12.2	11.0	10.0	9.2	8.5	7.9	7.3	6.9	6.1	5.5
120	30.0	24.0	20.0	17.1	15.0	13.3	12.0	10.9	10.0	9.2	8.6	8.0	7.5	6.7	6.0
130	32.5	26.0	21.7	18.6	16.3	14.4	13.0	11.8	10.8	10.0	9.3	8.7	8.1	7.2	6.5
140	35.0	28.0	23.3	20.0	17.5	15.6	14.0	12.7	11.7	10.8	10.0	9.3	8.8	7.8	7.0
150	37.5	30.0	25.0	21.4	18.8	16.7	15.0	13.6	12.5	11.5	10.7	10.0	9.4	8.3	7.5
160	40.0	32.0	26.7	22.9	20.0	17.8	16.0	14.5	13.3	12.3	11.4	10.7	10.0	8.9	8.0
170	42.5	34.0	28.3	24.3	21.3	18.9	17.0	15.5	14.2	13.1	12.1	11.3	10.6	9.4	8.5
180	45.0	36.0	30.0	25.7	22.5	20.0	18.0	16.4	15.0	13.8	12.9	12.0	11.3	10.0	9.0
190	47.5	38.0	31.7	27.1	23.8	21.1	19.0	17.3	15.8	14.6	13.6	12.7	11.9	10.6	9.5
200	50.0	40.0	33.3	28.6	25.0	22.2	20.0	18.2	16.7	15.4	14.3	13.3	12.5	11.1	10.0

Row Spacing determines length of 1/100th of an acre
 10.0 feet row spacing = 43.5 feet = 1/100th of an acre
 9.5 feet = 45.9 feet = 1/100th of an acre
 9.0 feet = 48.4 feet = 1/100th of an acre
 8.5 feet = 51.2 feet = 1/100th of an acre
 8.0 feet = 54.45 feet = 1/100th of an acre
 7.5 feet = 58.1 feet = 1/100th of an acre

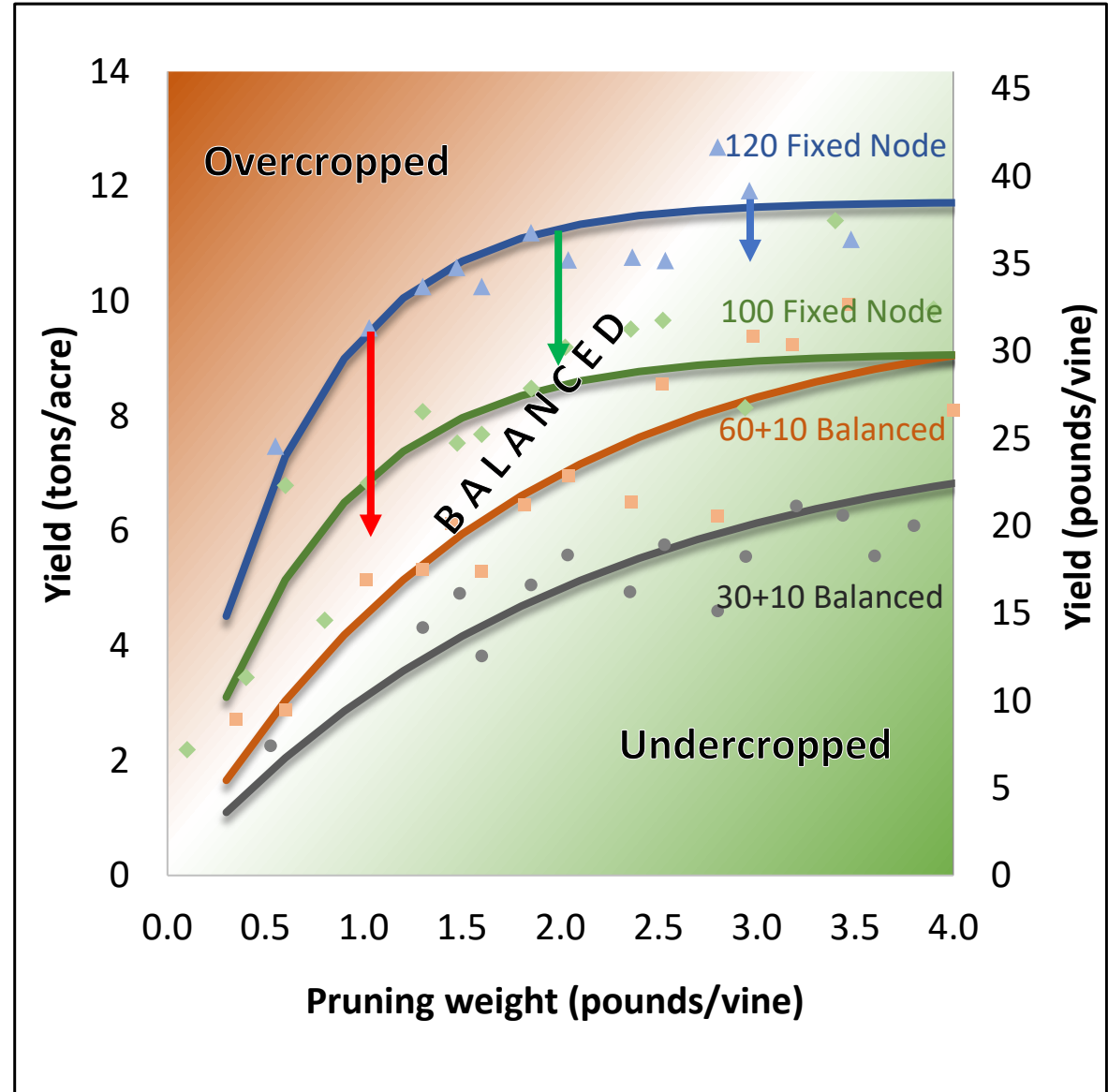
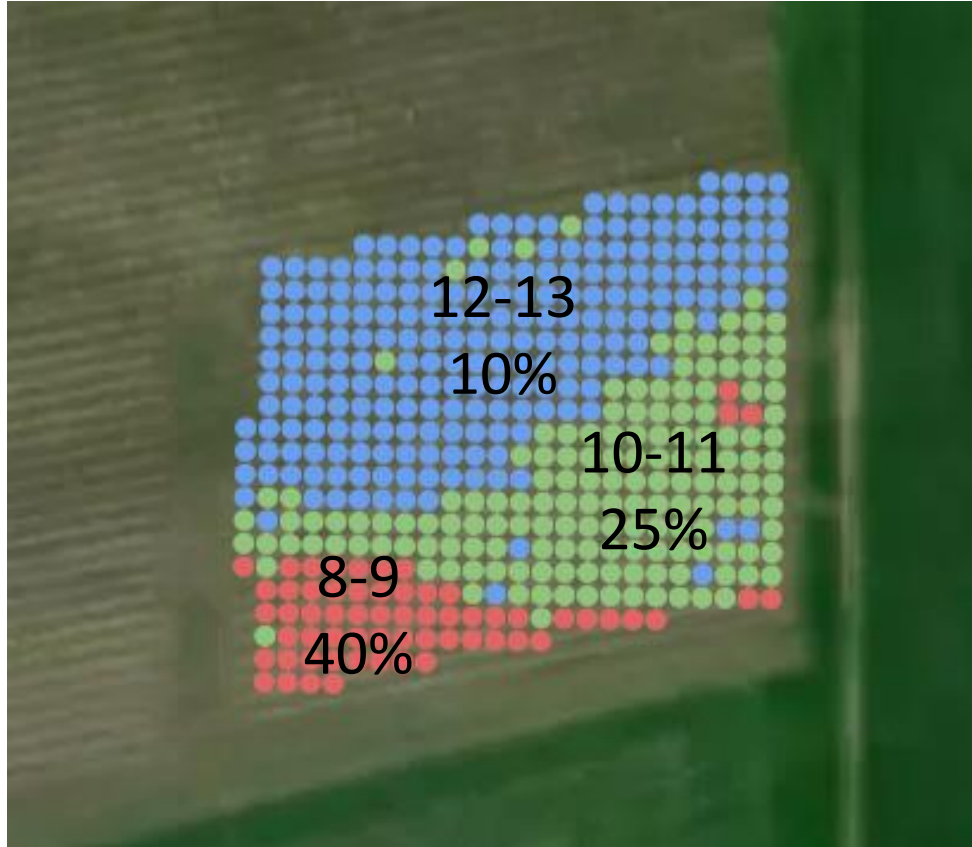
Calculation
 43, 560 square feet per acre
 Divide by row spacing and then
 divide by 100 to get 1/100th of an acre

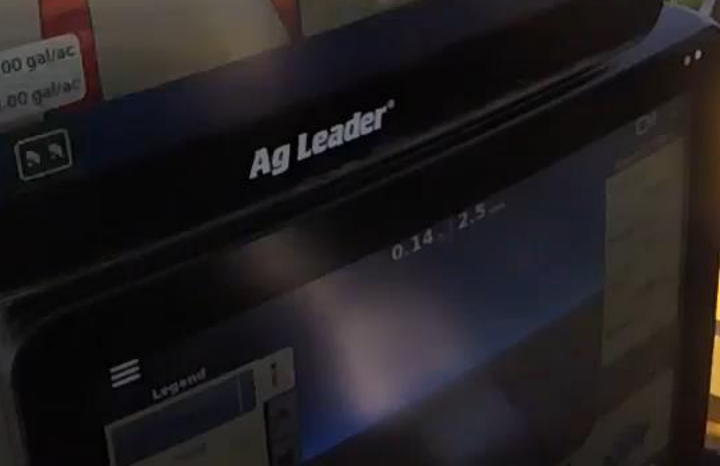
Example:

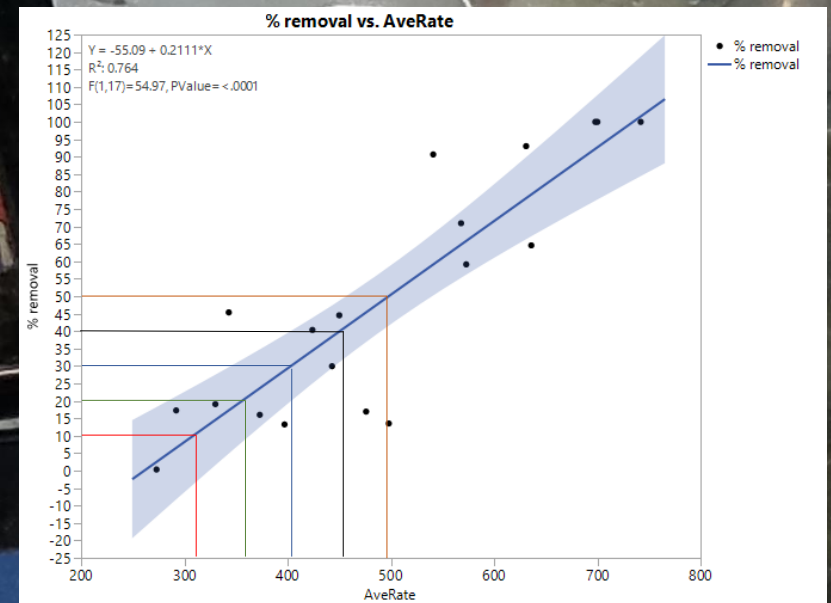
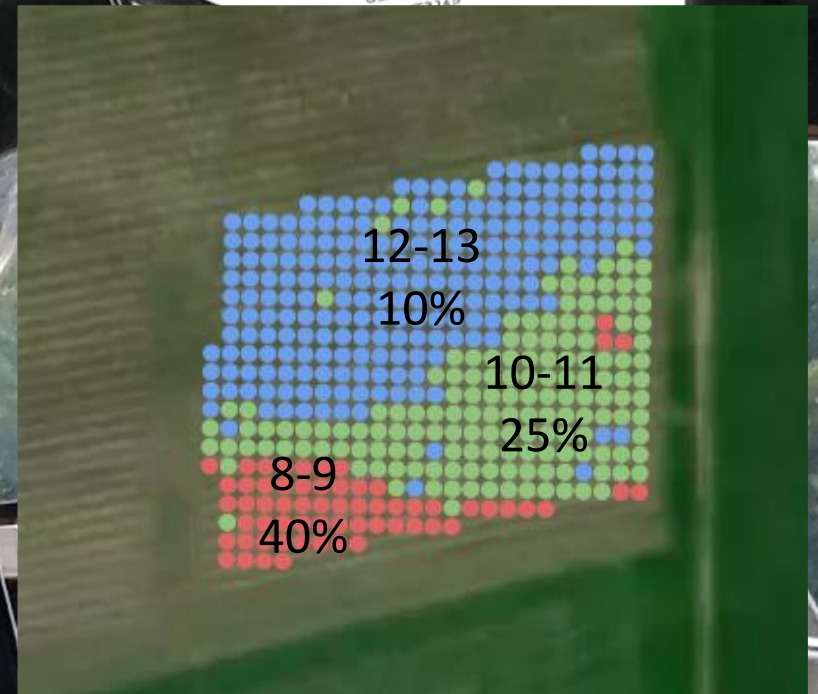
A grower has 9 foot row spacing and clean picks 48.4 feet at 25 days after bloom. The fruit weighs 80 pounds and the grower estimates that the berries are between 35% and 40% of final berry weight. According to the table, the crop estimate is between 10.0 and 11.4 tons per acre.

Disclaimer:

This table gives the relationship between time of season and % final berry weight on an average year. Year to year variability in weather related berry growth adds error to this table. Information on current year berry growth can be obtained from the Fredonia Vineyard Lab (or) it is strongly suggested that individual growers start collecting berry weight information from their own individual vineyard blocks.

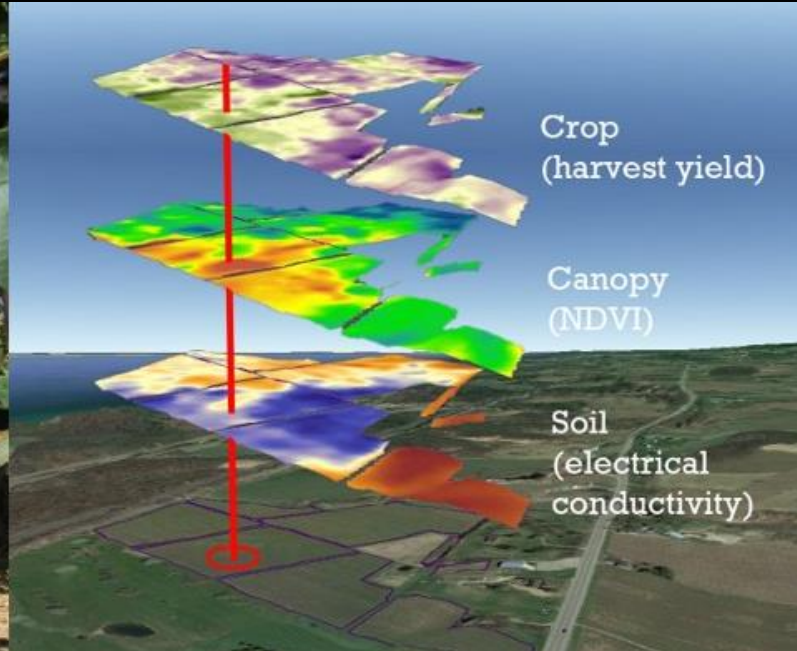








The Efficient Vineyard Approach

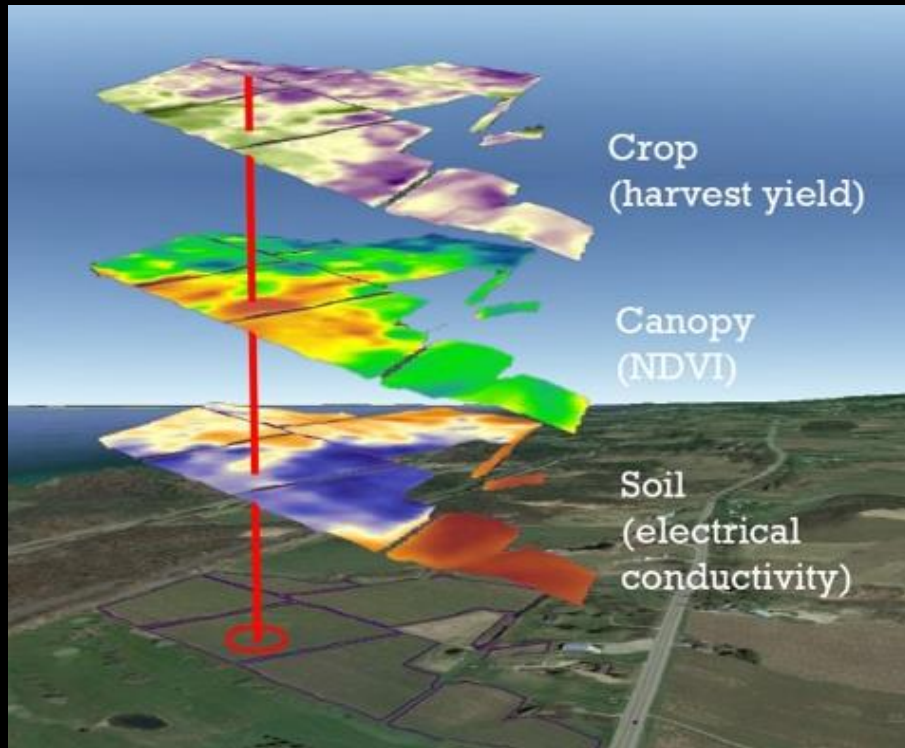


Measure vineyard soil, canopy, and crop characteristics using mobile field sensors

Model multi-layer spatial data needed for perennial cropping systems

Manage vineyards by integrating spatial information with variable-rate technology

Development of Efficient Vineyard Software (aka: The MyEV Tool)



Spatial Data Barrier

We found that processing and working with spatial information was a barrier for growers to adopt new precision viticulture management techniques.

Why not use existing platforms?

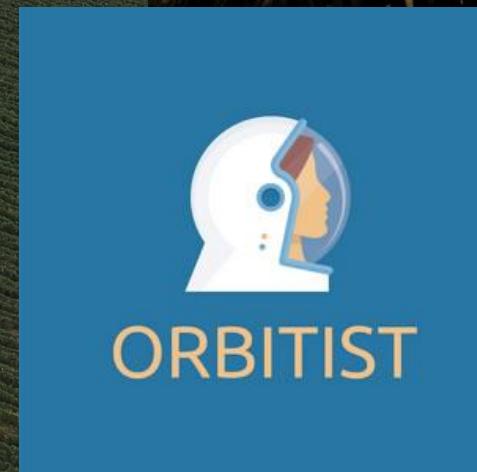
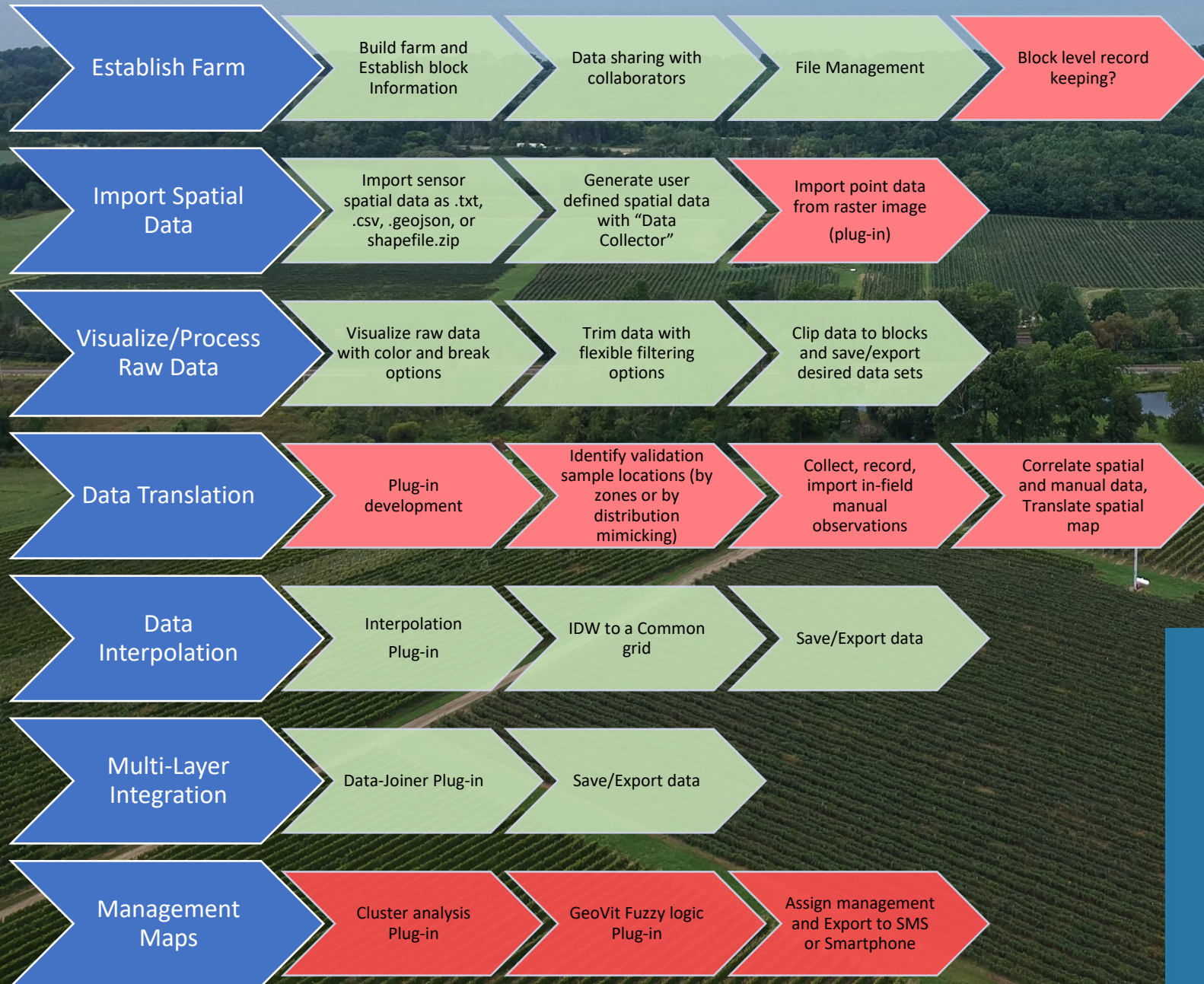
Powerful GIS software, such as ArcGIS and QGIS, provide many functions but have a substantial learning curve and are not agriculture specific.

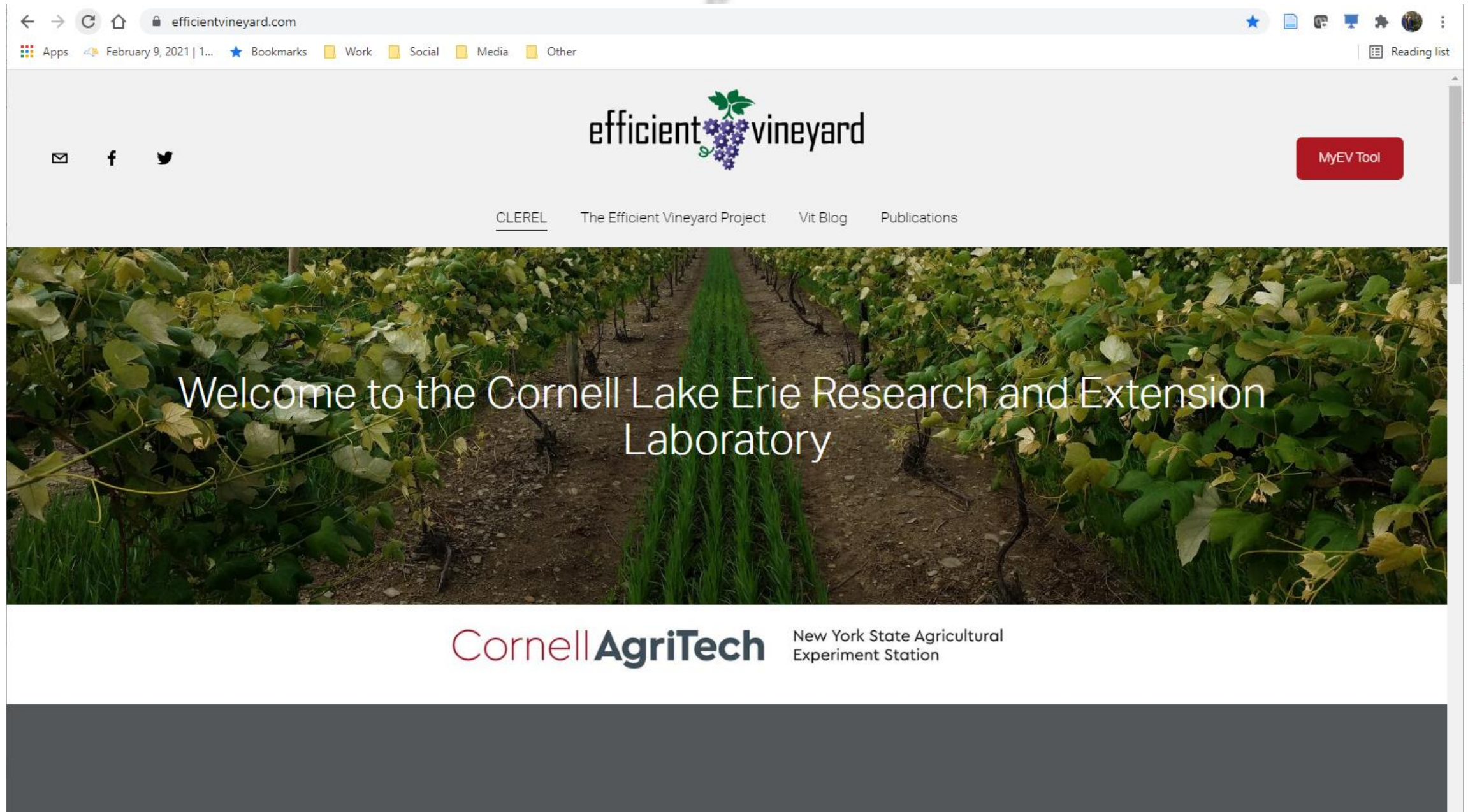
The MyEV goal

Our goal was to develop a simple, easy to use, web-based software tool for vineyard managers to upload, process, visualize, and download spatial data.

Who did this work

MyEV is a cooperative effort between Terry Bates (Cornell Viticulturist) and Nick Gunner (Software Developer and owner of Orbitist, <https://orbitist.com/>).





Dr. Terry Bates, Cornell AgriTech, Lake Erie Research and Extension Laboratory



Email

you@email.com

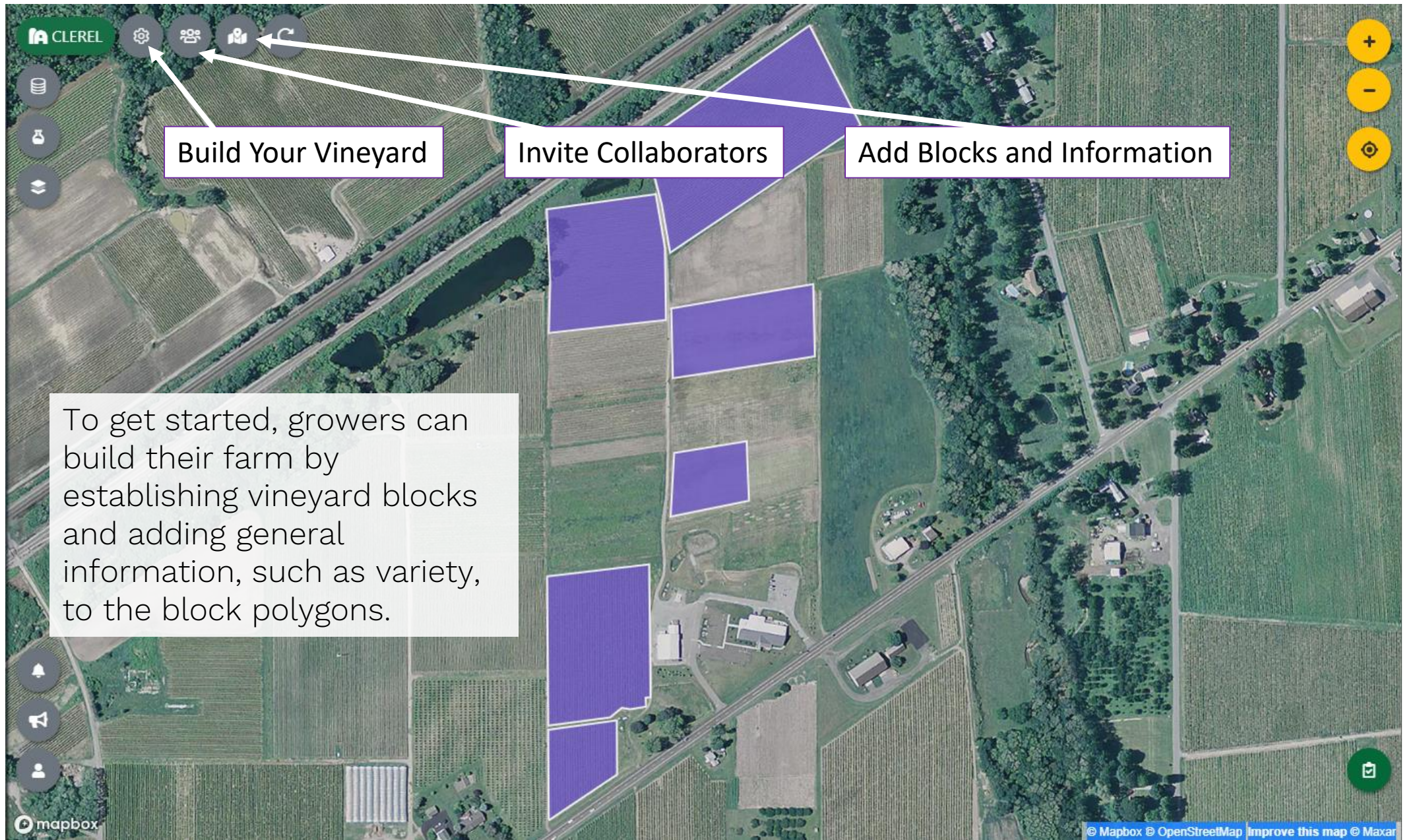
Password

Log In

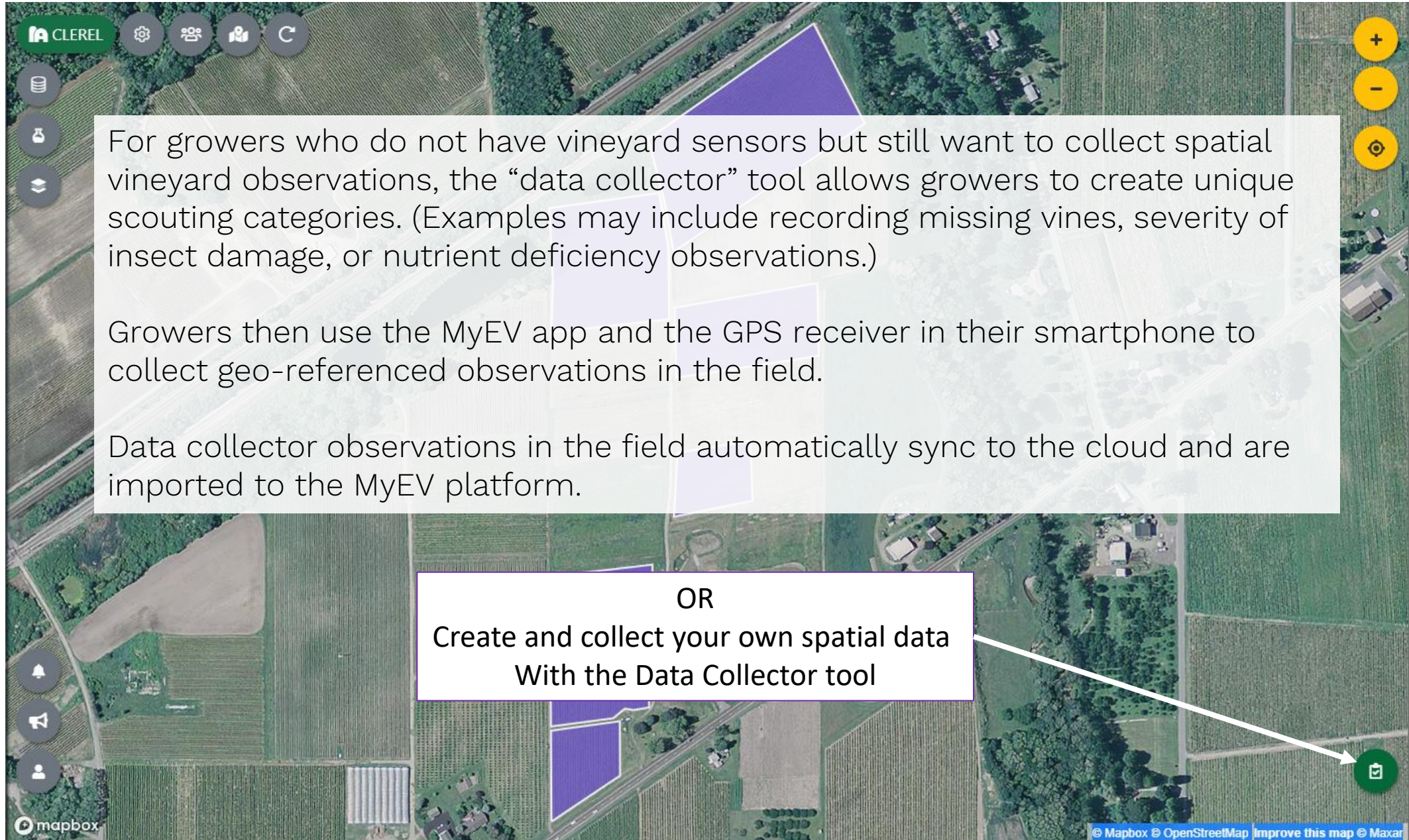
[Forgot Password](#)

[Create an Account](#)

First time users are directed to set up a free account and password.







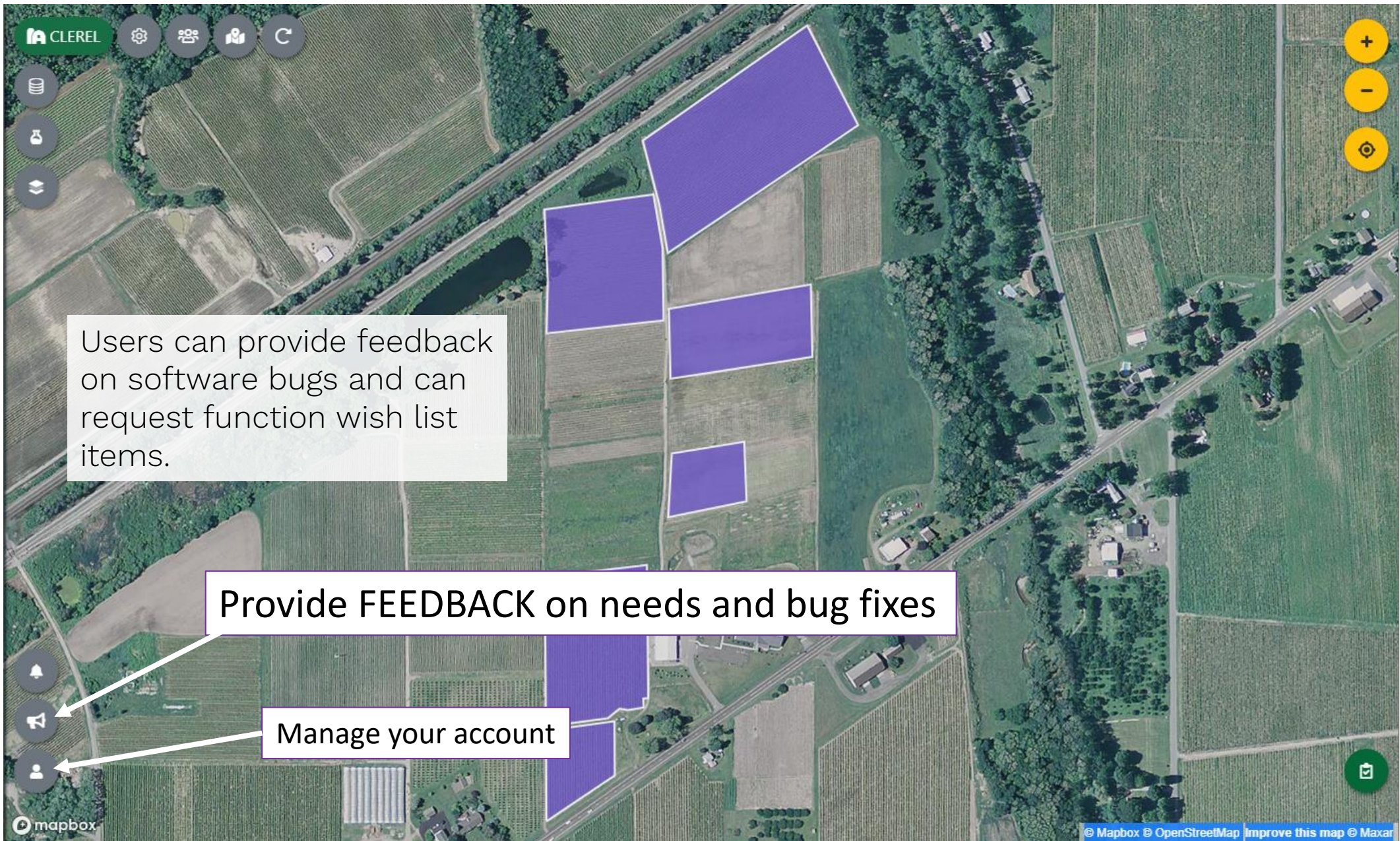
For growers who do not have vineyard sensors but still want to collect spatial vineyard observations, the “data collector” tool allows growers to create unique scouting categories. (Examples may include recording missing vines, severity of insect damage, or nutrient deficiency observations.)

Growers then use the MyEV app and the GPS receiver in their smartphone to collect geo-referenced observations in the field.

Data collector observations in the field automatically sync to the cloud and are imported to the MyEV platform.

OR

Create and collect your own spatial data
With the Data Collector tool



Users can provide feedback on software bugs and can request function wish list items.

Provide FEEDBACK on needs and bug fixes

Manage your account

Upload Dataset

Name

CLEREL_NDVI_June_09_2020

Description

Tags

raw 2020 NDVI CLEREL

Upload Data File

Can be a .txt, .csv, .geojson, or shapefile.zip file with 'LAT' and 'LNG' or 'Latitude' and 'Longitude' headers.

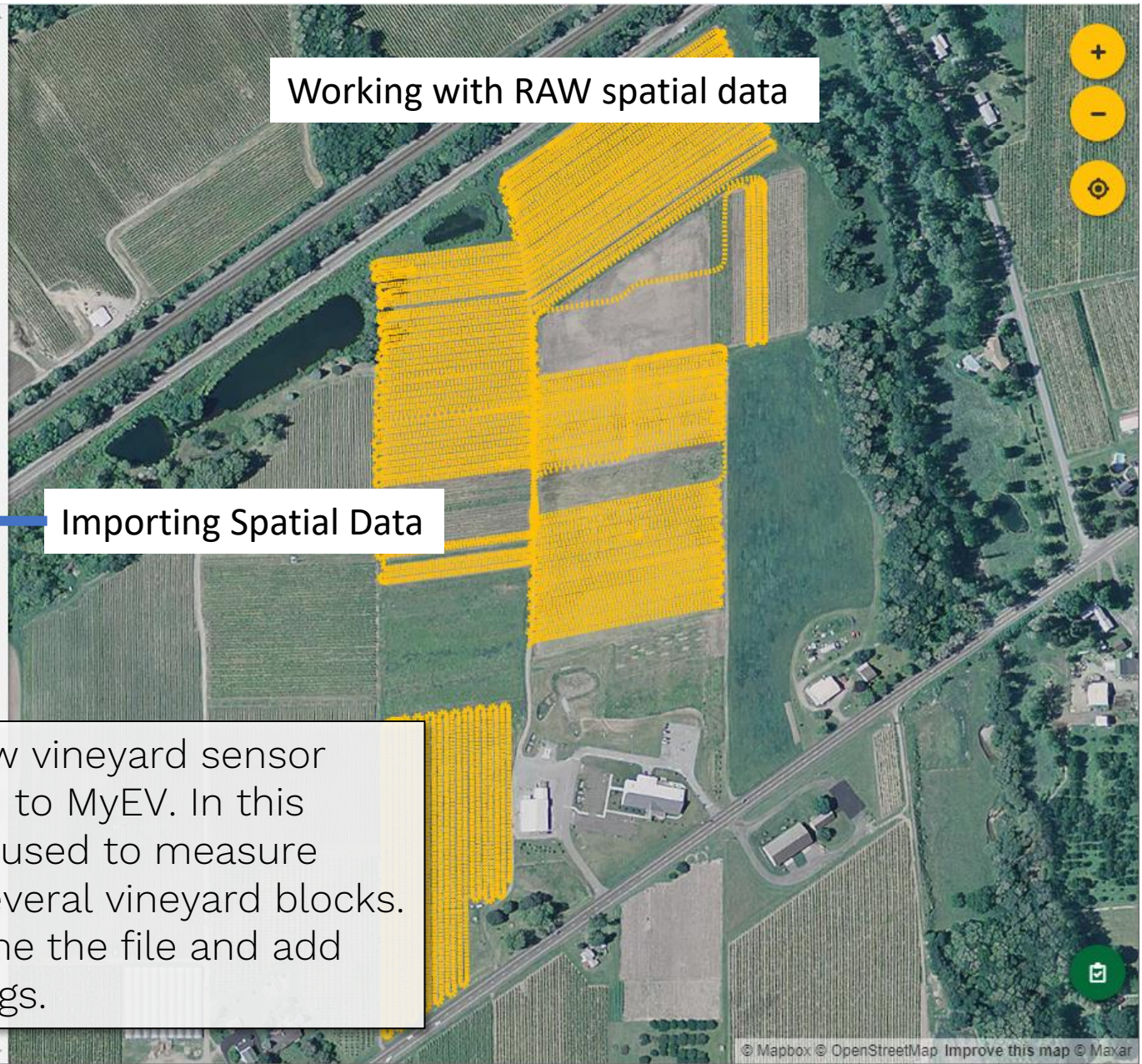
Choose File

CLEREL_NDVI_June09.2020.txt

CLEREL_NDVI_June09.2020.txt

Submit

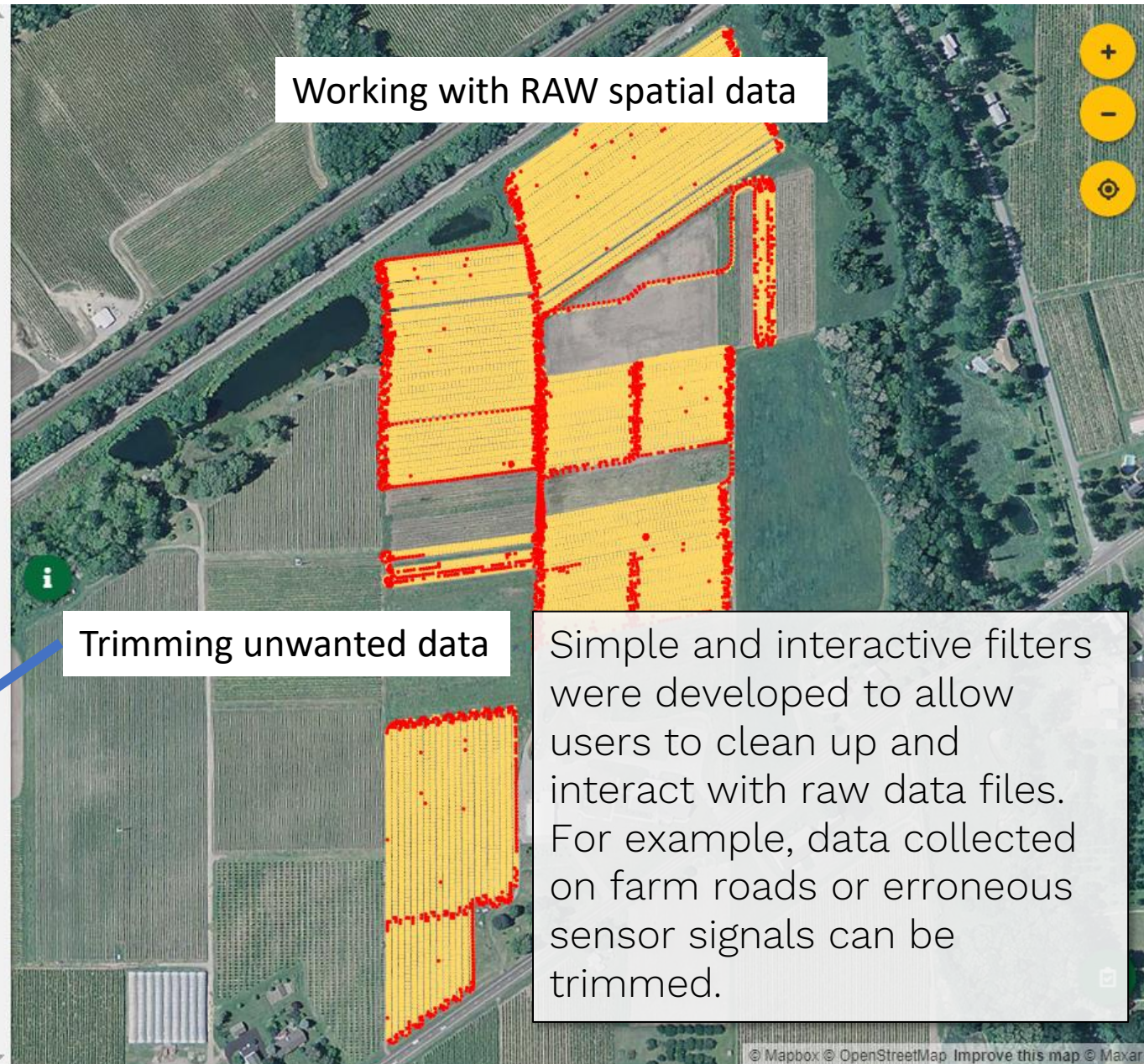
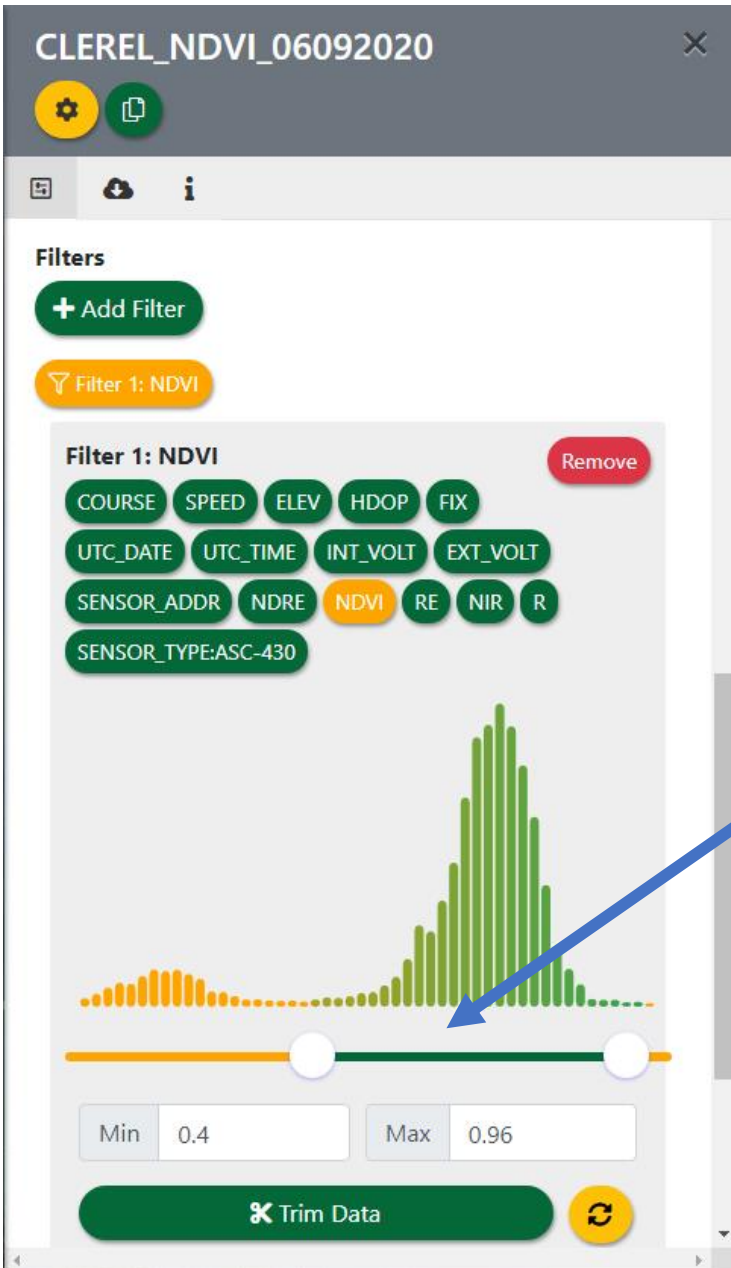
Cancel



Working with RAW spatial data

Importing Spatial Data

This is an example of what raw vineyard sensor data looks like when imported to MyEV. In this example, an NDVI sensor was used to measure grapevine canopy growth in several vineyard blocks. On import, the user can rename the file and add descriptive information and tags.



CLEREL_NDVI_06092020

Map Type

Point Map

Heat Map

Color Setting

☐ Solid

☒ Value Based

Color by Header

Ramp Settings

Linear

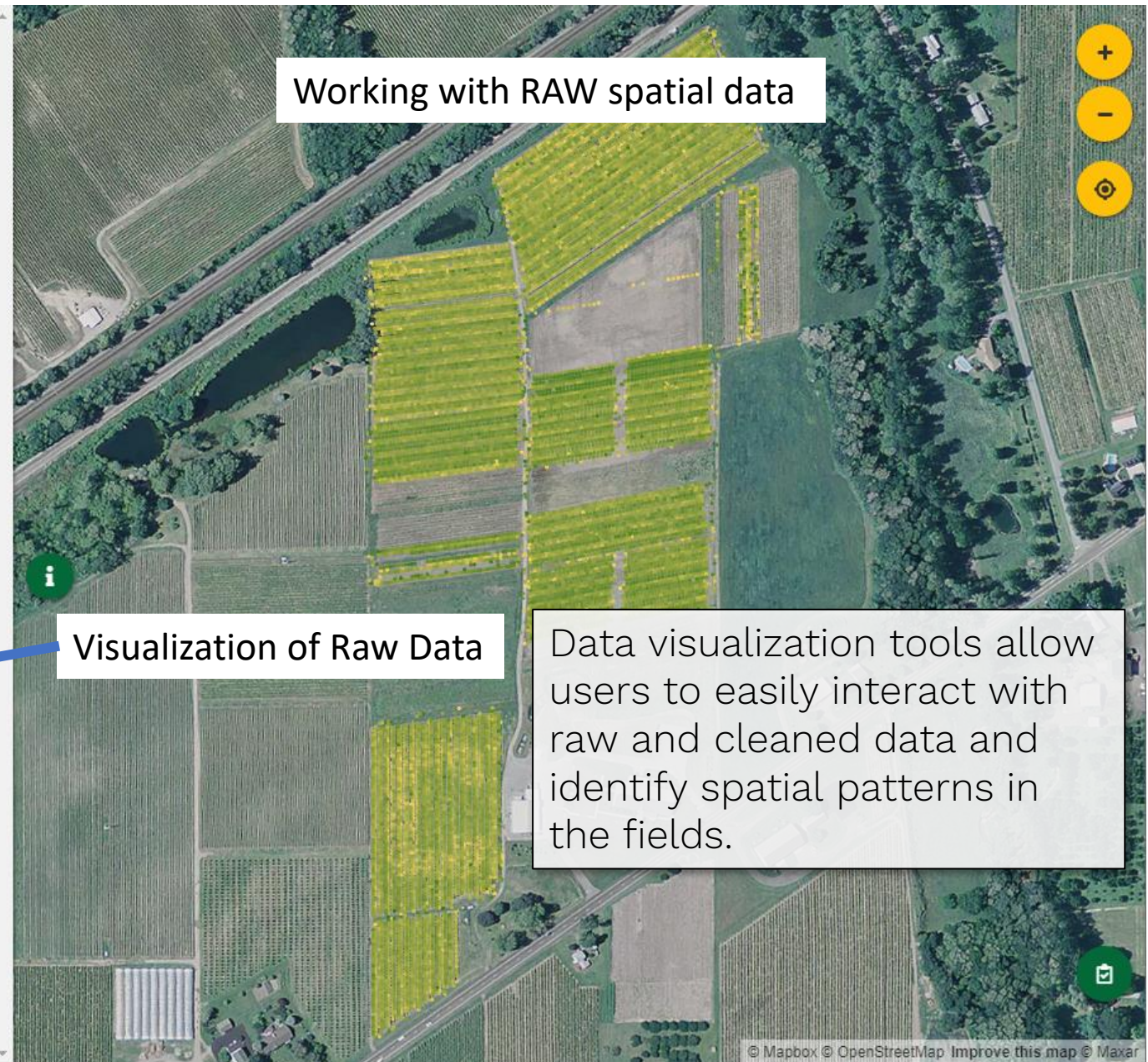
Min/Max Colors

Point Size Setting

☒ Fixed

☐ Value Based

Point Size: 2



Data Point

COURSE: 80.5

SPEED: 6.3

ELEV: 229.36

HDOP: 0.7

FIX: GPS

UTC_DATE: 90620

UTC_TIME: 172750

INT_VOLT: 4.18

EXT_VOLT: 13.59

SENSOR_ADDR: 2

NDRE: 0.237

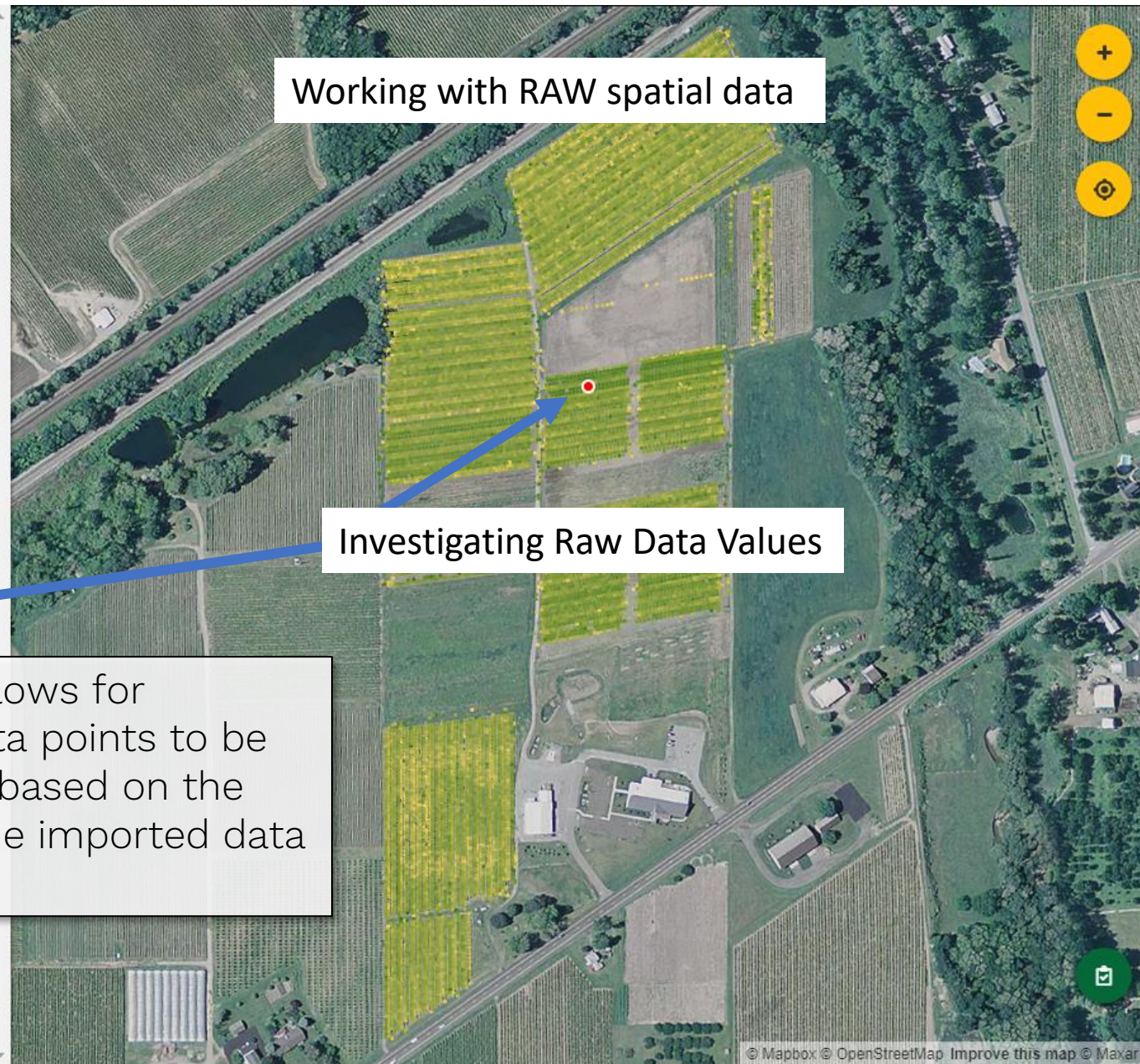
NDVI: 0.857

RE: 20.573

NIR: 33.362

R: 2.572

SENSOR_TYPE:ASC-430: null



Copy Dataset

New Dataset Name

Copy_CLEREL_NDVI_06092020

☒ Filter data before copying

This will remove datapoints based on any filters you have set. Leave this unchecked if you want to copy the entire dataset.

☒ Trim data using selected blocks

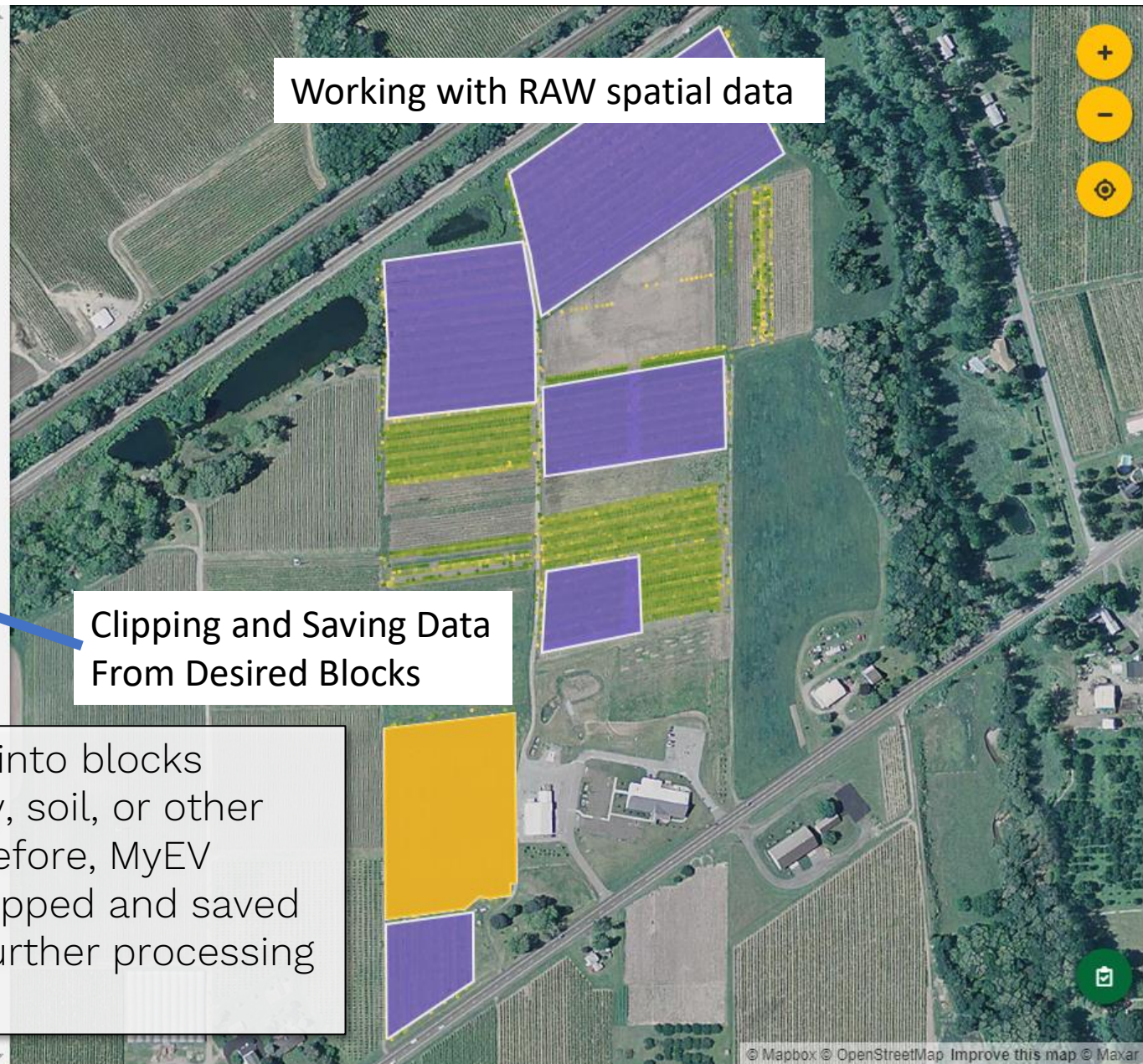
This will remove datapoints outside of blocks you select on the map. Leave this unchecked if you want to copy the entire dataset.

Selected Blocks:

- Barn Block
- Barn Block

Copy Dataset

Cancel



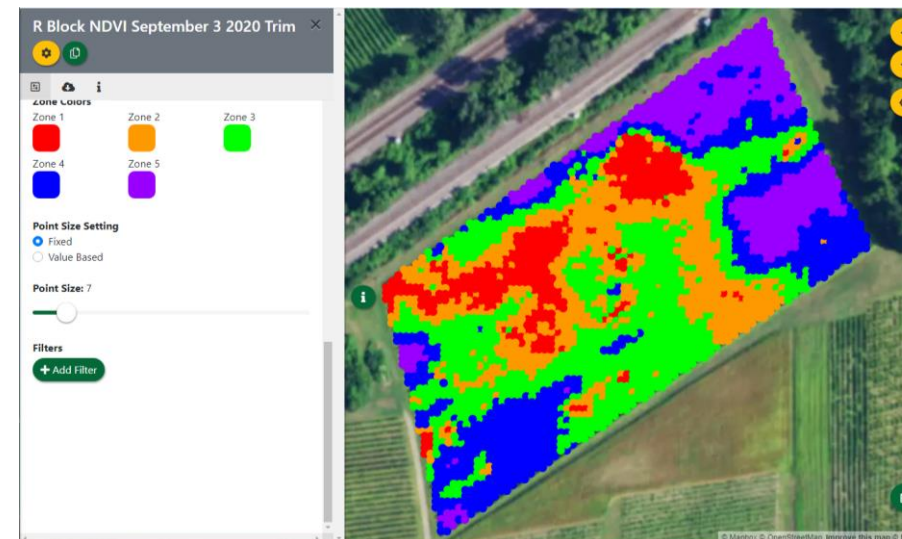
Most vineyards are broken up into blocks based on differences in variety, soil, or other management parameter. Therefore, MyEV allows large data files to be clipped and saved to individual farm blocks for further processing and management.



Proximal field sensors in the vineyard

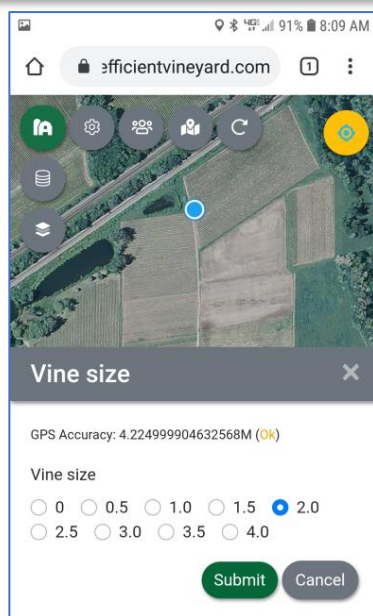
High End Investment

Vineyard Sensors
External GPS
Commercial Data Logger



High density spatial data from proximal field sensor

MyEV is flexible in accepting geo-referenced data from field sensors or the MyEV data collector



MyEV Smartphone app showing data collector

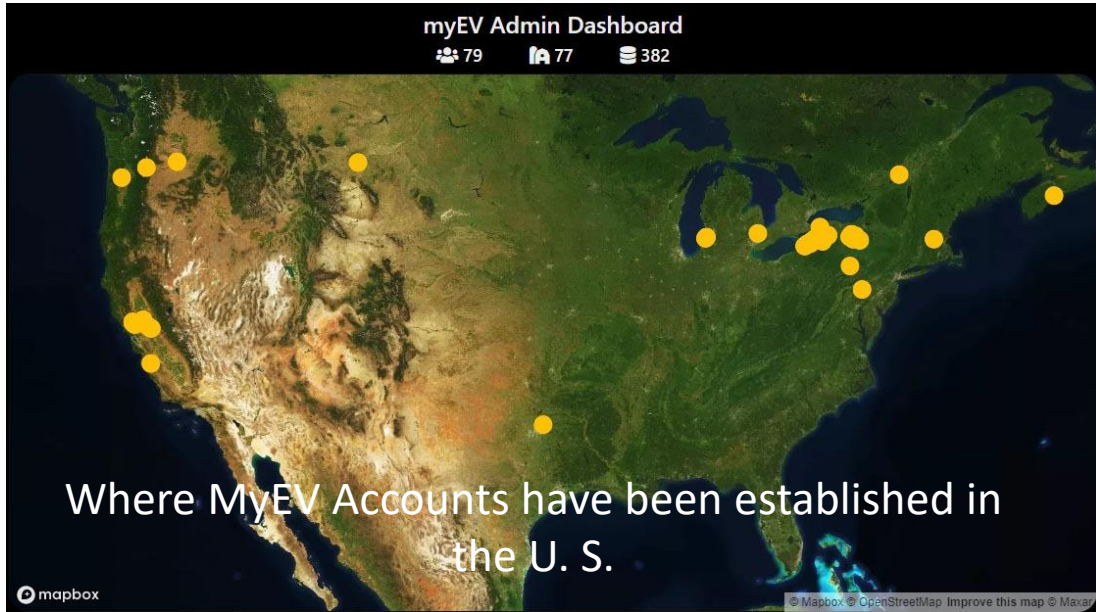
Low End Investment

Human Observation
GPS enabled phone
“Data Collector” function

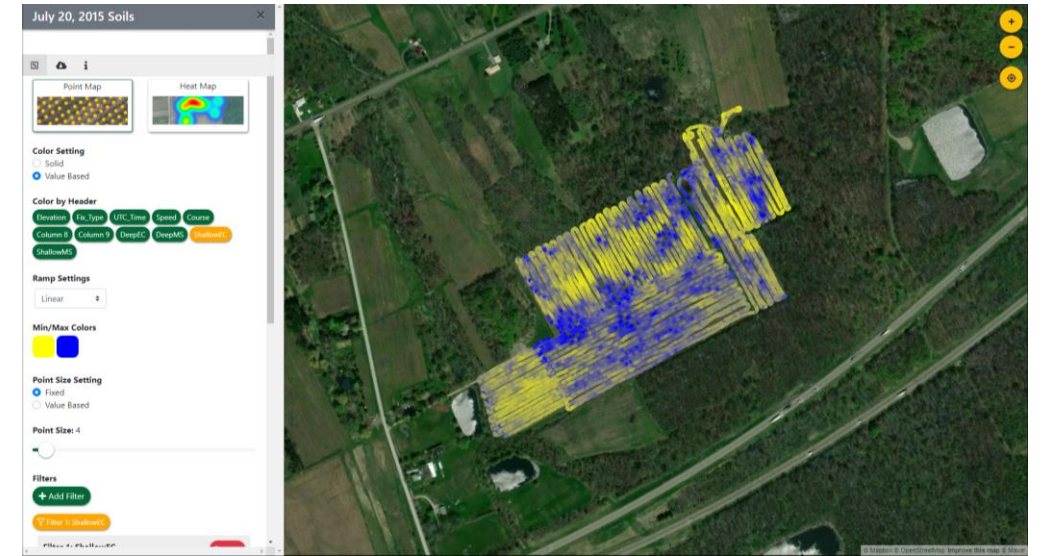


Moderate density spatial data from visual observations recorded on MyEV smartphone app

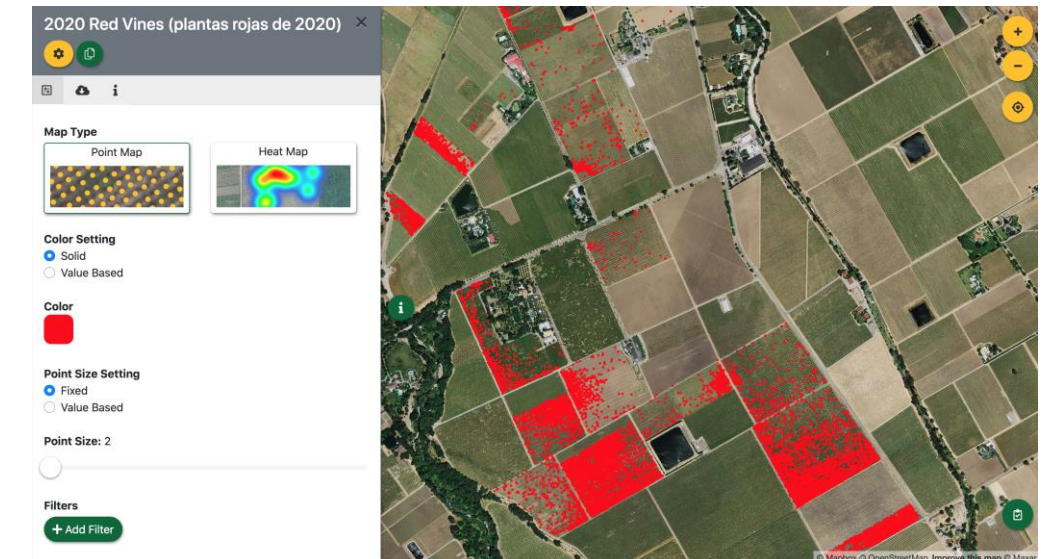
How and Where is MyEV being used by the Industry



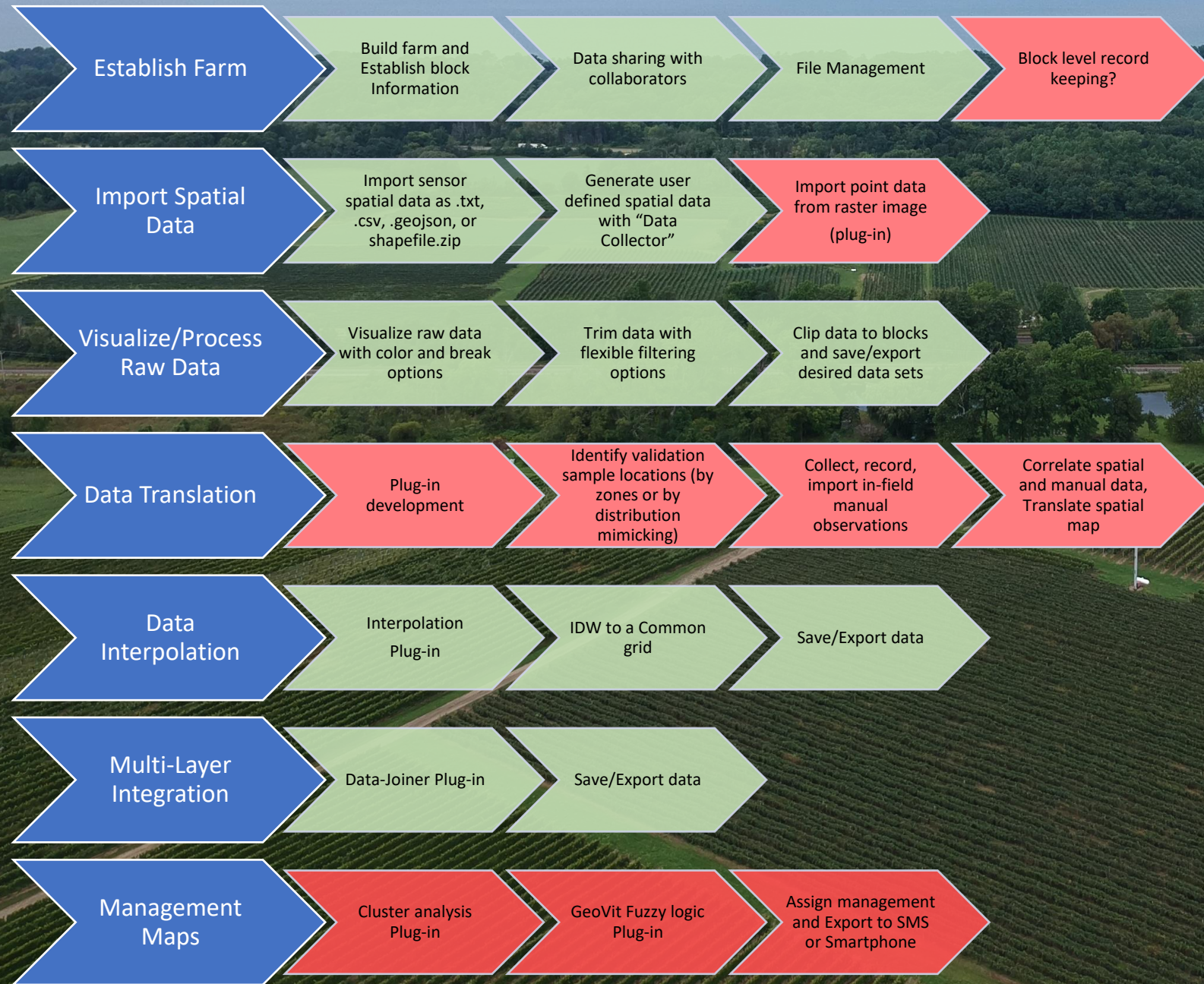
Where grapes are produced in the U. S.



Vineyard Soil mapping in New York



Scouting for virus infected vines in CA



Accomplishments and Planned Development for MyEV

The base MyEV platform for establishing user accounts and block maps, importing and visualizing raw data, and basic data interpolation and exporting has been developed and is currently live and accepting grower feedback.

Future development will focus on advanced plug-in development for data processing, in-field data validation, and generating spatial prescription management maps.