

# 11A: Introduction to data products from aerial imagery

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## Aerial imagery

1. Payloads
2. 3D reconstruction
3. Multispectral and hyperspectral imagery
4. LIDAR

In this lecture you will learn:

- Typical imaging payloads
- Data products

# Imaging systems are the most frequently used payloads on drones

- Imaging comes in many flavors
  - Passive systems:
    - Visual spectrum (electro-optical) camera
    - Multispectral
    - Hyperspectral
  - Active systems
    - LIDAR
      - Line scanning
      - Imaging
    - Structured light

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Parrot Sequoia  
Multispectral sensor



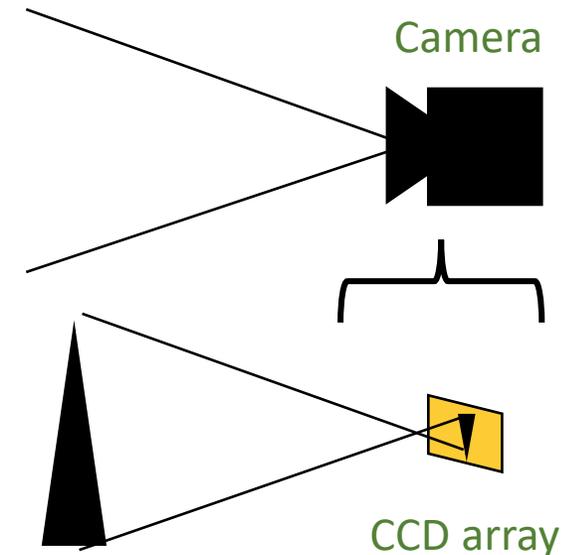
Bayspec hyperspectral imager

# Electro-Optical (EO) Cameras

- Electro-optical cameras work with a CCD (charge coupled device) or CMOS (complementary metal oxide sensor) sensor
  - These sensors consist of an array of semiconductor “pixels” to provide a full frame description of the visual field of view
  - The size of the pixels and the size of the array determines the size of the image, typically expressed in megapixels
  - CCD or CMOS cameras can be either global shutter or rolling shutter
  - CCD cameras tend to be cheaper but CMOS sensors are much more energy efficient

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# EO camera global vs. rolling shutter design

- Rolling shutter cameras work with a progressive line scan exposure which can cause distortion of moving objects
- Rolling shutter cameras are typically cheaper, with most CMOS-based cameras having a rolling shutter
- The rolling shutter design is usually acceptable for video capture but it is not acceptable if any frame-by-frame analysis is required (such as in 3D reconstruction or anomaly detection)

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The rotor of the Yamaha RMAX is not symmetrical in this image due to the rolling shutter 4

# EO camera global vs. rolling shutter design

- Global shutter cameras are required in any photogrammetry work or 3D terrain reconstruction when camera motion or motion on the ground can impact image quality
- Global shutter cameras are more expensive than rolling shutter cameras
- If all that is needed is video, or the scene is not dynamic then a rolling shutter camera will work

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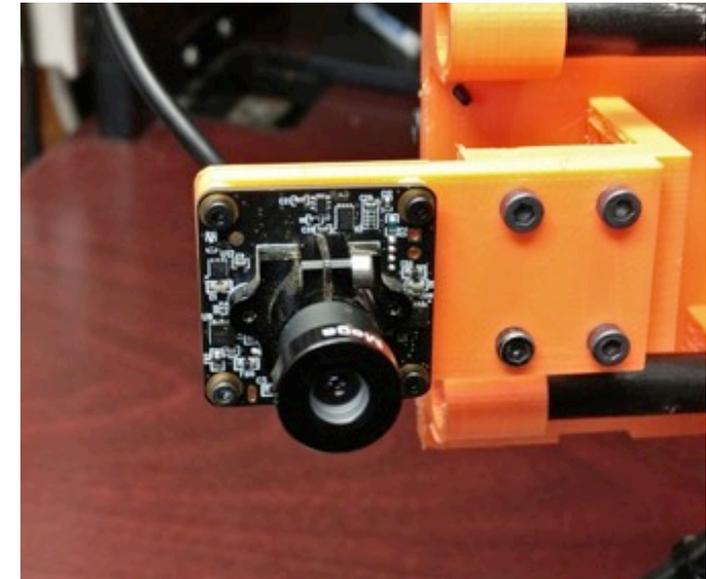
In this global shutter image of the RMAX, the rotor is symmetrical and shows no asymmetry

# Camera installations are frequently at “board level” to save weight

- In aircraft installations, cameras and other hardware may be considered “original equipment manufacturer” (OEM) equipment
  - These payloads do not have the packaging necessary for stand-alone operation
  - They are intended to be integrated into other systems
  - In the case of drones, this saves a great deal of weight and allows for a customizable design

## Aerial imagery

1. **Payloads**
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A camera mounted on PC board that is integrated into a custom drone application

# Larger aircraft can accommodate larger, stabilized camera payloads

- These are more expensive since they have very accurate pointing capability
- They can typically hold a stabilized view from several hundred meters distance
- The gimbal can be controlled by a joystick to change views

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VT RMAX carrying a stabilized camera payload on nose



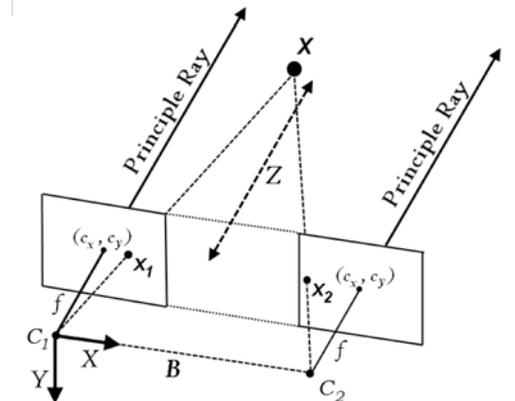
# Stereovision systems allow for a rapid 3D reconstruction of the scene

- A stereovision system matches corresponding features in the left and right cameras, and defines a disparity  $d$  between the two features in the images
  - This disparity corresponds to a  $Z$  distance from the cameras to the feature:
$$Z = \frac{fB}{x_1 - x_2} = \frac{fB}{d}$$
  - The collection of features with a disparity  $d$  can be represented in a disparity map



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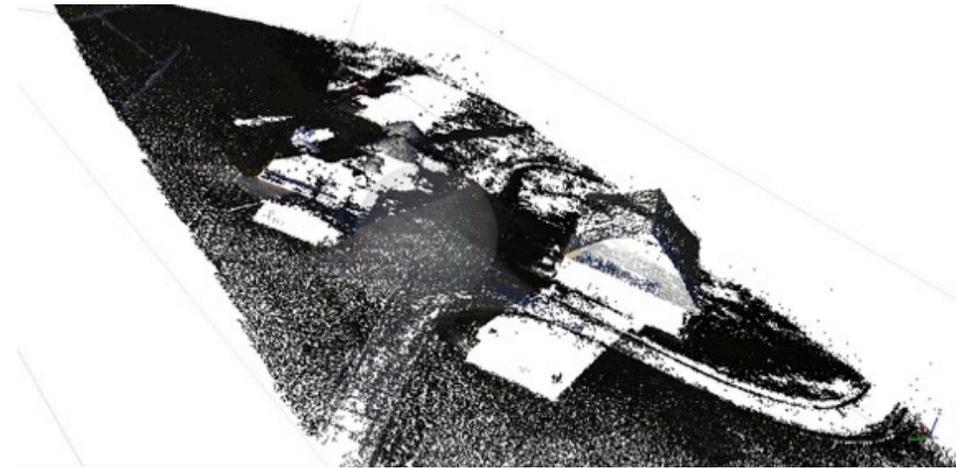


# A set of image frames over a scene can be used to reconstruct the 3D environment

- Building a 3D terrain model using “structure from motion” is a process that minimizes error of a set of 3D point locations that are described in multiple image frames
- The computational time to create the 3D terrain will increase with number of image frames, and the processing can take a very long time for large datasets (hundreds of images)
- The first step is to achieve correspondence between image frames

## Aerial imagery

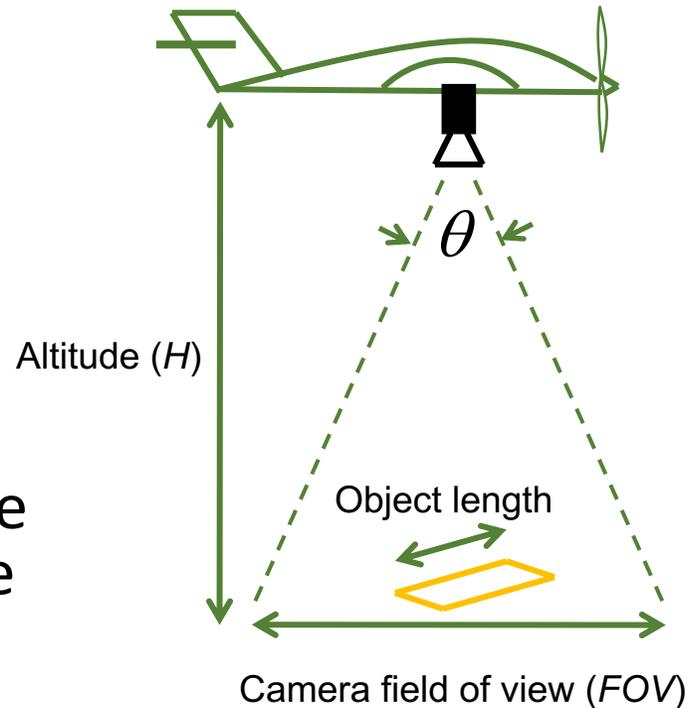
1. Payloads
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A 3D point cloud generated from a series of images taken from a single drone camera

# Actually, the first step in building a 3D reconstruction is collecting the images

- It is necessary to achieve a certain % overlap in all directions so that there is correspondence between neighboring images
- The field of view (FOV) of the camera is computed from the imaging angle and height of the aircraft
- For 75% overlap, can you write the relationship between speed of the aircraft and the frame rate of the camera?



$$\tan\left(\frac{\theta}{2}\right) = \frac{FOV}{2H}$$

*Note that a camera imaging angle of  $53^\circ$  corresponds to the  $FOV = H$  (a 1:1 relationship)*

## Aerial imagery

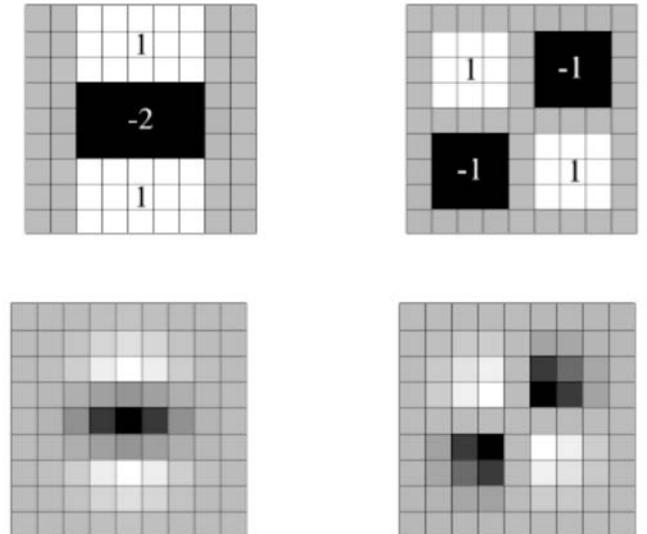
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# Aligning image frames

- In order to build a terrain model, registration of points between successive images is required
- There are many ways to do this, but it is common to define **unique features** in the image pixel-space for frame-to-frame matching
  - **SURF** features: Speeded Up Robust Features use a Hessian matrix-based detector and distribution-based descriptor
    - A Gaussian derivative of a binary kernel is used to detect unique features
  - Approximations of point correspondence reduces the space for feature matching

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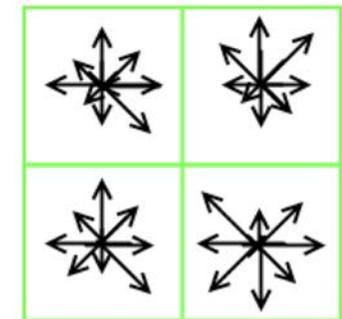
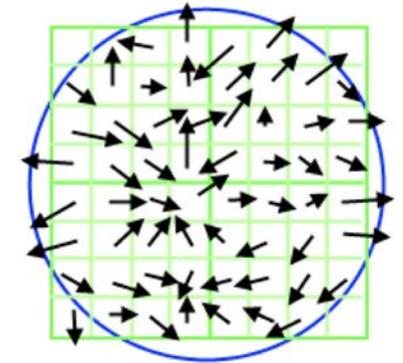
The Gaussian kernel is passed over the image to identify unique features in frame-to-frame matching

# Aligning image frames

- Another feature commonly used for alignment is the **SIFT** feature – scale invariant feature transform
- SIFT features define an orientation vector from the 2x2 Hessian kernel
  - The vector is based on the max and min response orientation of the feature
  - The descriptor consists a 4x4 array with eight orientations, resulting in a 128-element descriptor, unique to scale

## Aerial imagery

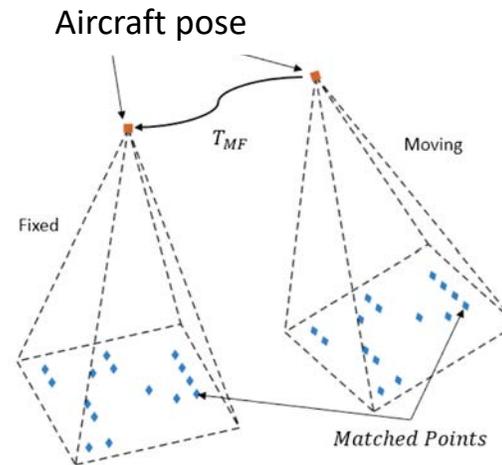
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A vector description  
is scale invariant

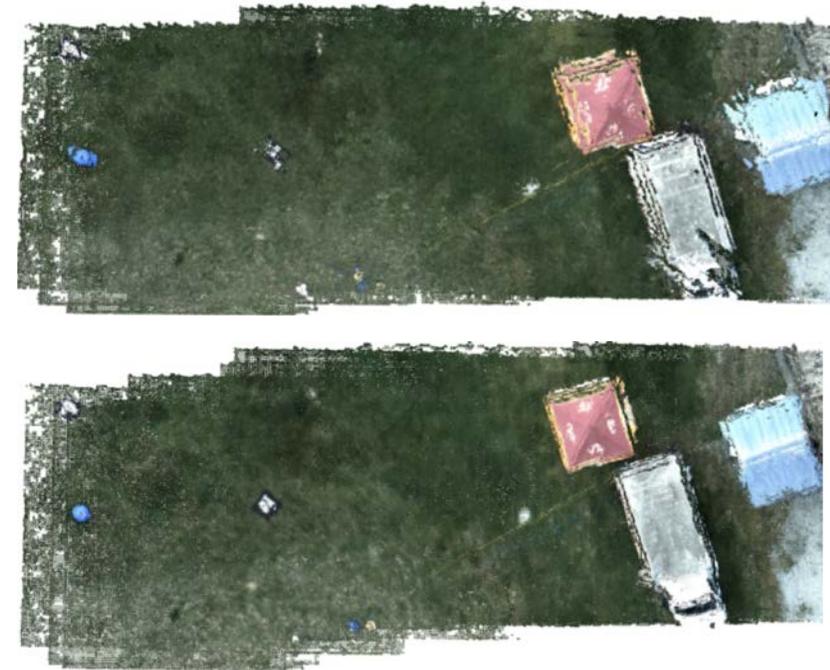
# Aligning image frames

- Frame-to-frame correspondence is approximately accomplished by using the IMU data from the aircraft
  - The pose of the aircraft is reasonably well defined by the aircraft GPS, accelerometers and gyros
- Final matching of frames is accomplished by feature point detection and matching



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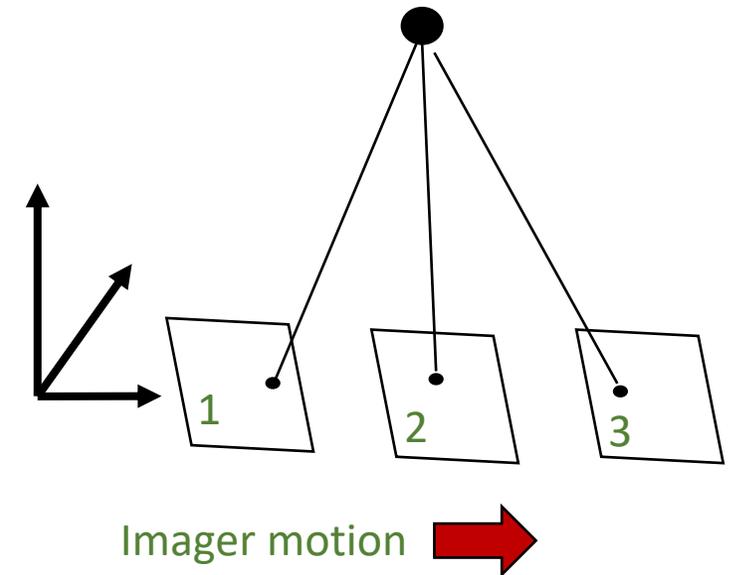
The top image shows a sequence of image frames w/o matching. The bottom image uses feature matching

# The 3D terrain is reconstructed by minimizing re-projected error

- Minimization of the error in point correspondence using camera pose is called bundle adjustment
- The least error reprojection of an observed point from the image plane over all images requires a non-linear least squares minimization
  - The Levenberg-Marquardt algorithm is frequently used to accomplish this non-linear least squares minimization

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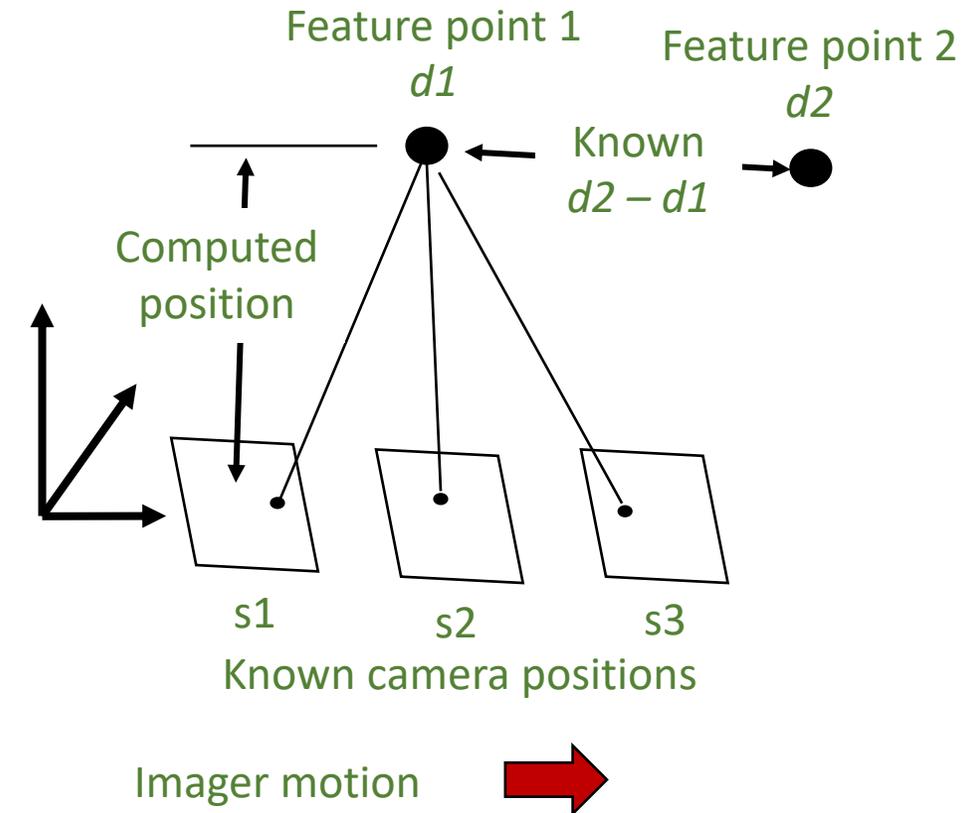
As the camera moves, the image plane point moves as a function of the camera pose relative to the point

# The resulting 3D reconstruction will have no scale information unless measurements are available

- If the **position of the camera is known** via GPS, then the observed relative motion of the terrain will be scaled properly
  - That will allow an accurate distance from the camera to be computed which results in scale feature point position
- If the **distance between features** in the scene (usually three point positions should be referenced) is known, then the entire scene can be properly scaled

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# The feature points used in reconstruction are a basis for full scene reconstruction

- Feature points used in 3D reconstruction represent good features for matching
  - They do not represent all of the pixels in the scene
- Dense point cloud reconstruction is completed using methods that fill in the remaining 3D description of the environment
  - The result is a 3D position for every pixel that can be related to the feature points
- In some cases, missing feature points caused by poor lighting or poor texture will result in holes in the 3D reconstruction

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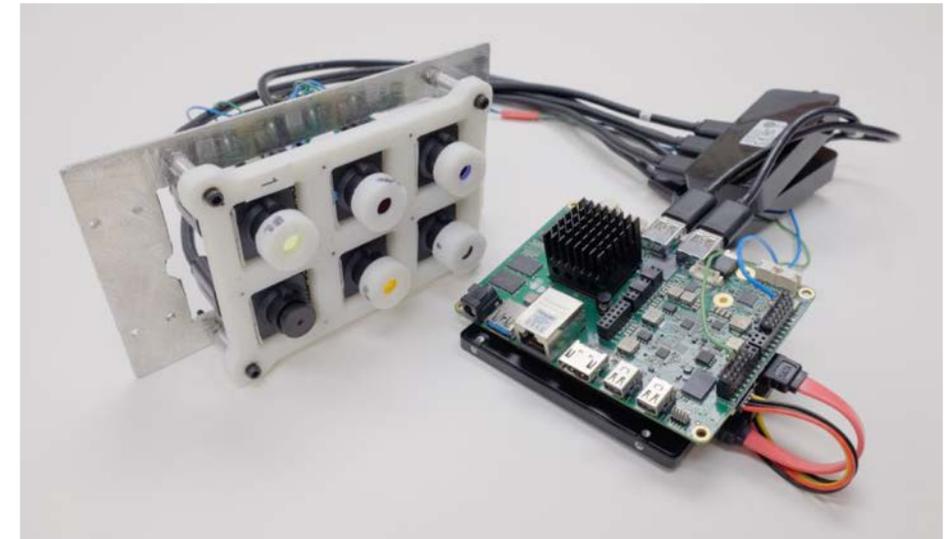
This scene has missing points due to poor texture in the roof of the building on the left

# Multispectral and hyperspectral data products are commonly used to find important environmental characteristics

- A **multispectral camera** is capable of filtering the image to show only certain EM bands of interest
  - These bands provide detailed information about crop health (biomass), moisture (wetlands), and classification for plant identification, etc
- A common index for biomass detection is the **Normalized Difference Vegetative Index (NDVI)** which is based on both near infrared and visual spectral bands

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Multispectral camera developed at the USL

$$NDVI = \frac{(R_{NIR} - R_{Red})}{(R_{NIR} + R_{Red})}$$

$R_{NIR}$  = NIR reflectance,  
 $R_{Red}$  = Red reflectance

# Multispectral imagery results have been correlated to crop yield data

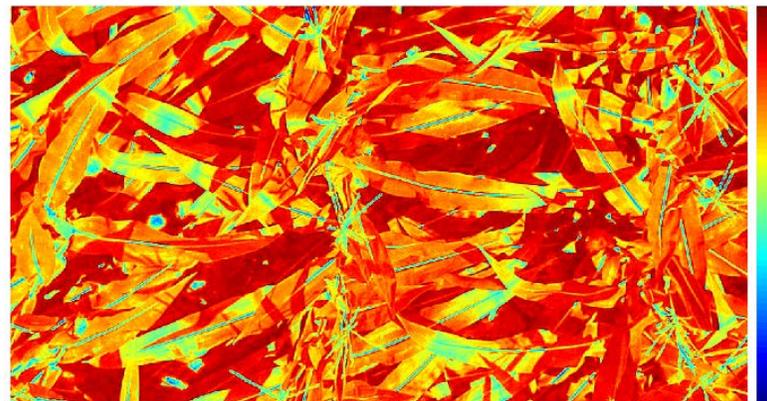
## Aerial imagery

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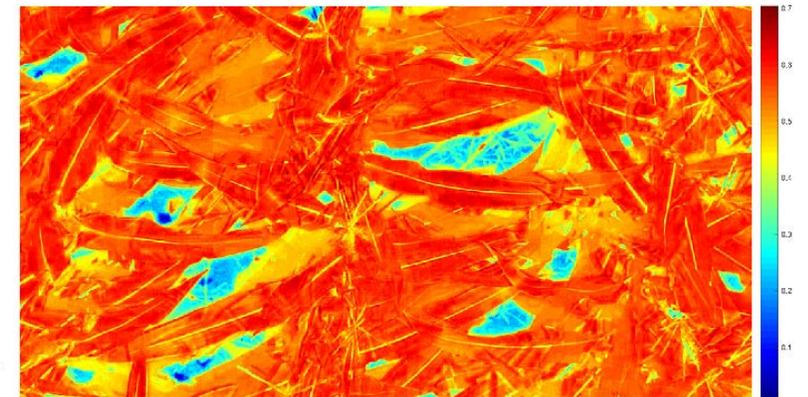
- Image processing should consider subtraction of the non-crop areas to get an accurate assessment of the crop index



Raw image of corn



NDVI of raw image



NDVI of image with background taken out

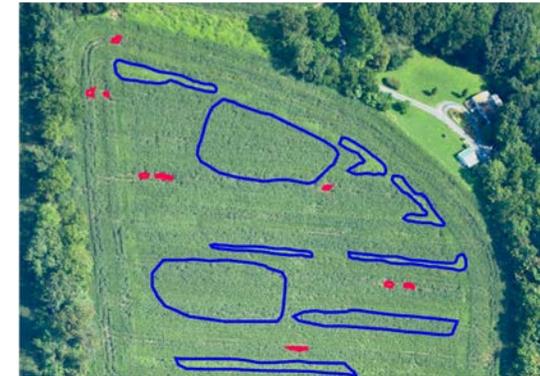
# It is possible to use visual spectrum data only to determine crop health or identify other agricultural features

- **Training using annotated ground truth data** creates a successful classifier for crop condition monitoring
- **RGB, HSV, or L-a-b** color spaces can be evaluated for best results
- Extracted features from the scene, like geometric shapes that are identified using the Hough transform (for example), can also be investigated in the visual spectrum

## Aerial imagery

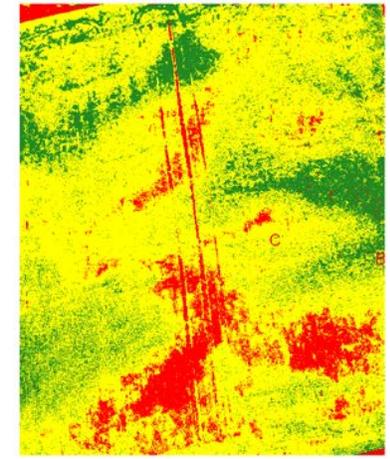
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Areas that are ground-truthed for a particular condition



Original

Classification



# Hyperspectral imagery

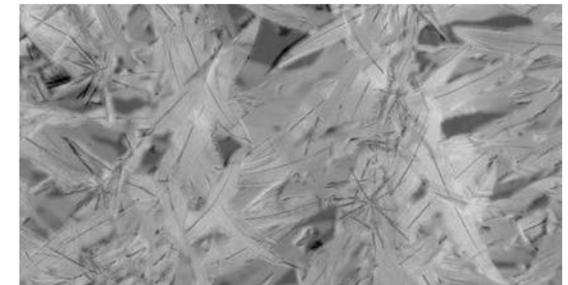
- **Hyperspectral imagery** uses narrow-band data to extract characteristics in the data that would normally not be identifiable
- Hyperspectral imagery from a typical sensor consists of a range of wavelengths from 600 nm – 1000 nm (Near IR), with 20 nm resolution
- Many hyperspectral imagers are “pushbroom” sensors, meaning that they function as line scanners and the camera must move to create a 3D image



OCI-UAV-2000  
hyperspectral camera  
(not a line scanner but  
a 256 x 256 imager)

## Aerial imagery

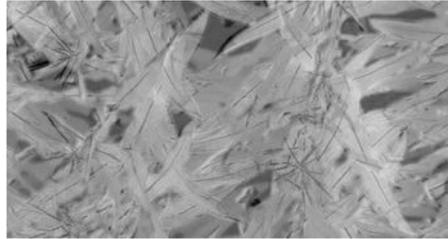
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Top: RGB image  
Bottom: 850 nm image

# Hyperspectral imagery

- **Creating indices of reflectance at certain bands** can reveal particular vegetative conditions that may allow targeted treatments or responses to disease

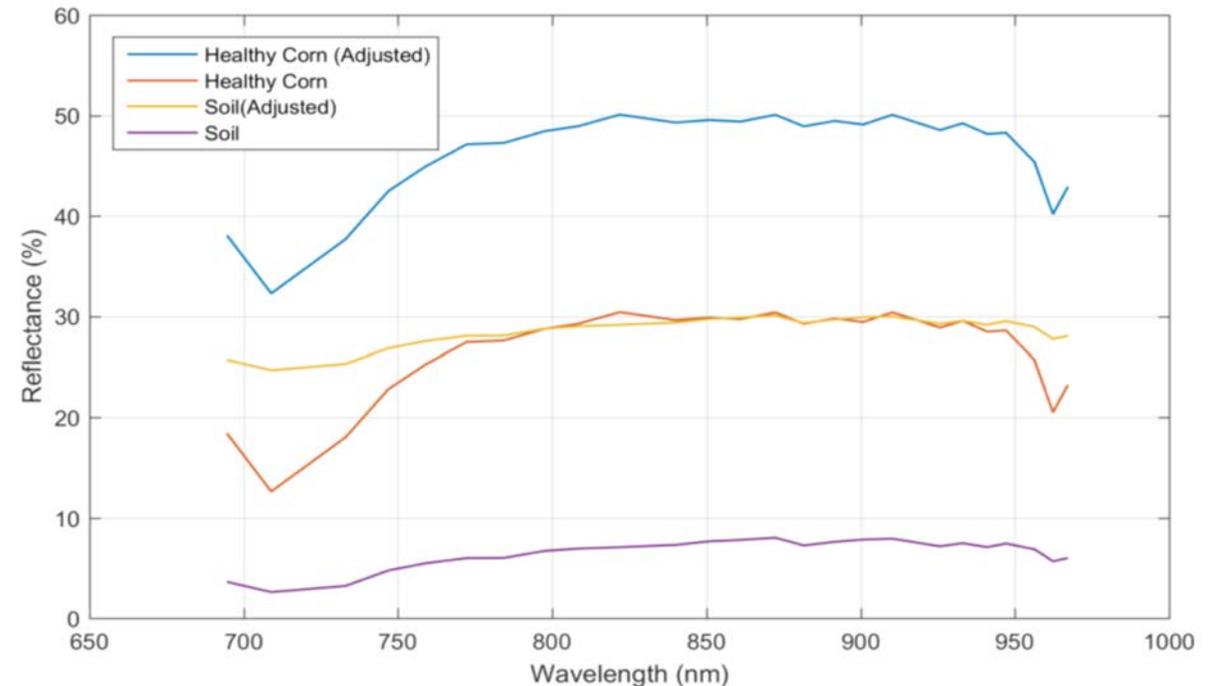


Aerial imagery

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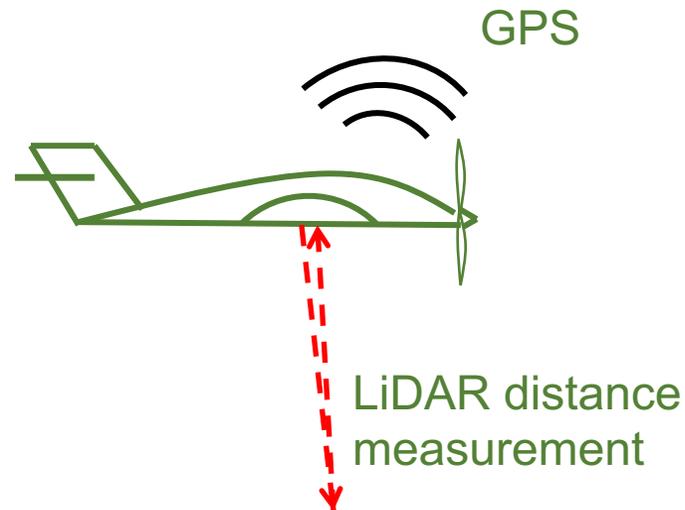
$$\text{Normalized difference Red Edge} = \frac{R_{790} - R_{720}}{R_{790} + R_{720}}$$

$$\text{Normalized difference Red Edge} = \frac{48 - 33}{48 + 33} = 0.185$$



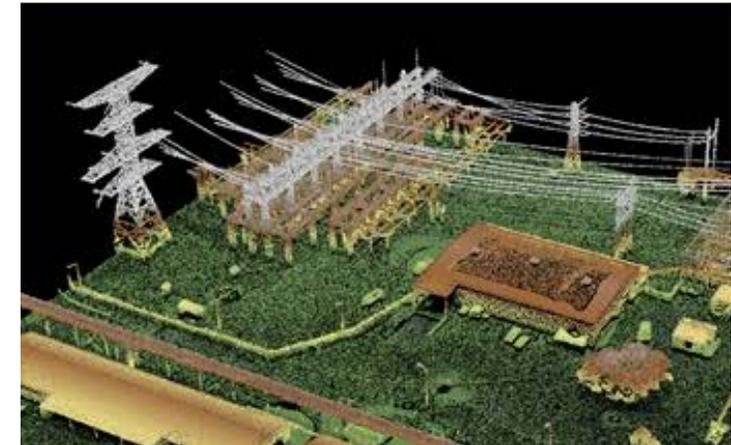
# LIDAR = Light Detection And Ranging

- A LIDAR system is an active sensor that sends a pulse of infrared energy to the ground, and it is reflected back up to the sensor
- The time of flight of the beam is used to estimate the distance from the sensor to the object that has been interrogated
- Accuracy and reliability tends to be higher than with vision – based systems, but the resolution is usually lower



## Aerial imagery

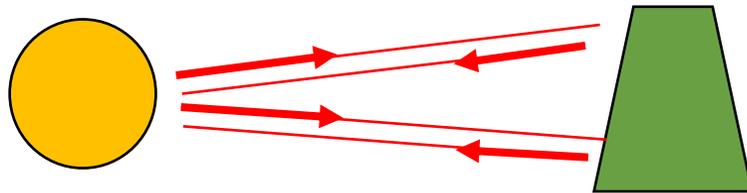
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Point cloud image from LIDAR

# LIDAR systems come in a variety of form factors which accommodate aircraft of many sizes

- Most line scanners emit a rotating IR laser beam with a timed detection system to measure time of flight
- The resolution is a function of the pulse time and the speed of rotation of a mirror that directs the beam in a 360 degree scan



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A Velodyne VLP-16 Puck LITE  
16 channel scanner with 100m  
range – suitable for drones

