

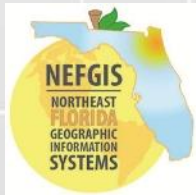
GIS Tools for Non-GIS Applications

Prepared for:

NEFGIS User's Group

11 April 2013

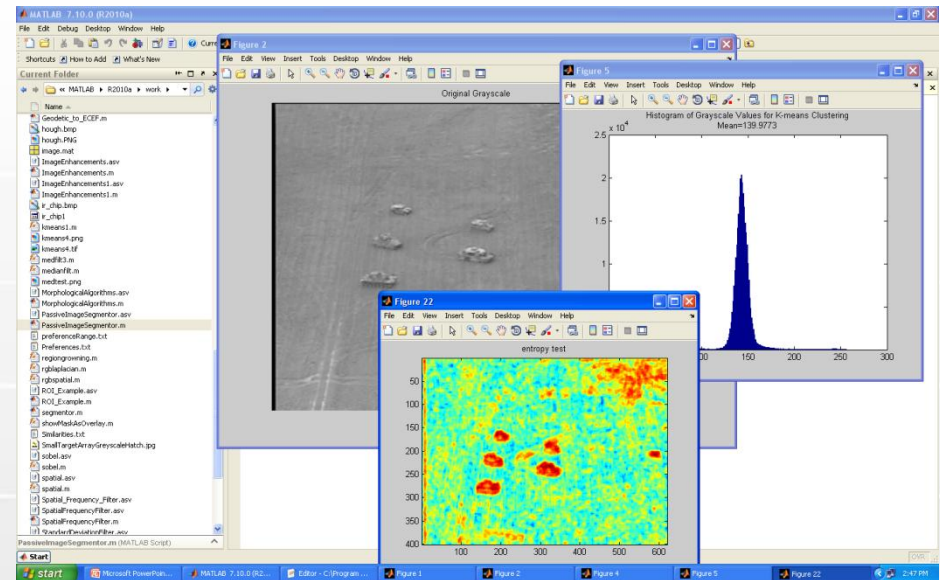
**Kevin Shortelle
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Overview

- Illustrate use of commonly-used GIS tools for non-GIS applications
- Introduce tool → GIS Application → Non-GIS Application
- Tools considered
 - Supervised Classification
 - Unsupervised Classification
 - Georeferencing
 - Spatial Filtering
 - Lidar Point Cloud Processing
- “No math, just business”

Steve Alter



Tools

- ➔ • **Supervised Classification**
- **Unsupervised Classification**
- **Georeferencing**
- **Spatial Filtering**
 - Low Pass (smoothing, averaging)
 - High Pass (sharpening)
 - Edge Detection
- **Lidar Point Cloud Processing**

Supervised Classification

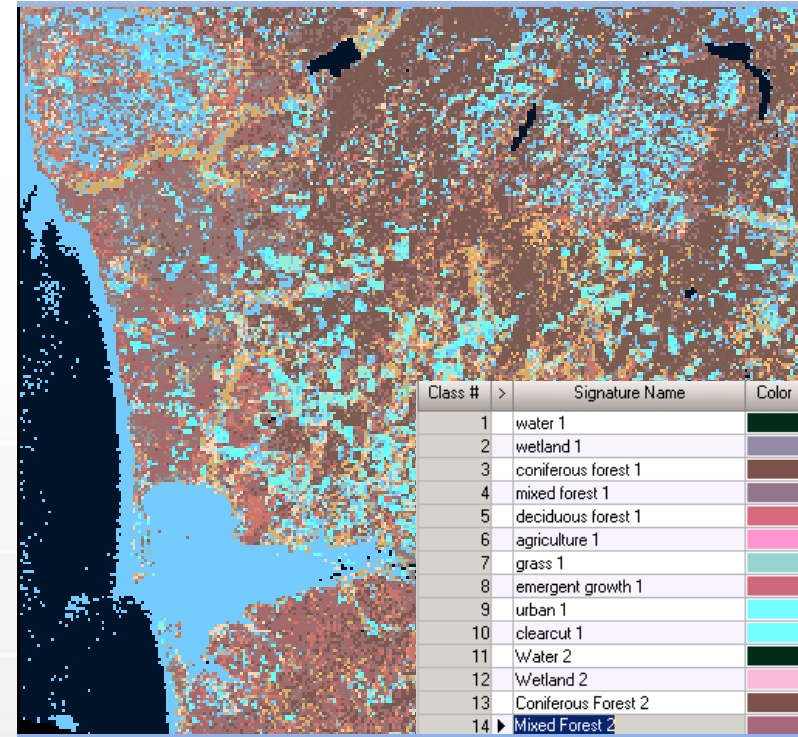
- **Analyst assigns selected raster pixels into known feature class**
 - Selected pixels are representative samples of features to be rendered in classified raster
 - Known feature classes referred to as training data or signature file
- **Classification algorithm uses training data to classify all remaining pixels in image**
 - Classification based spectral similarity between remaining pixels and known feature classes
- **GIS Tool**
 - ArcMap **Create Signature File** and **Maximum Likelihood Classification**
 - ERDAS Image **Signature Editor** and **Supervised Classification**

Supervised Classification (GIS)

Original Image



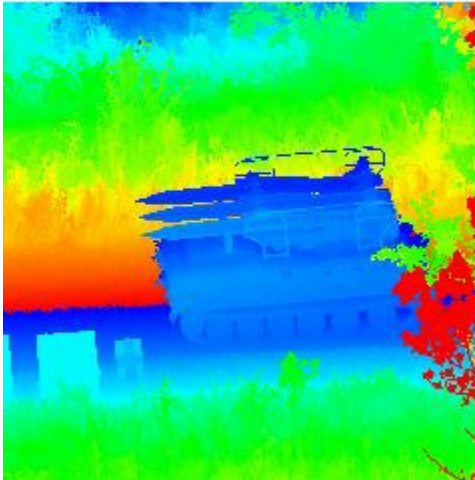
Classified Image



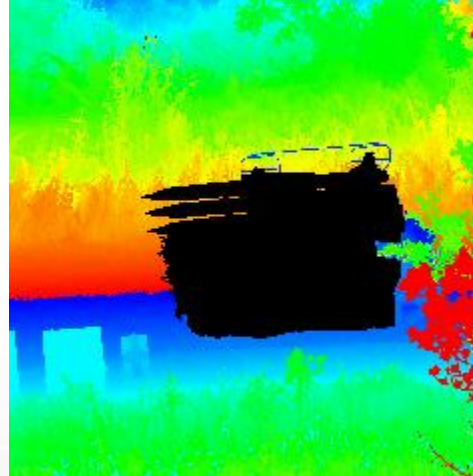
- Identify/select small areas of interest (AOI) in image that represent distinct features found across entire image (e.g., LULC application)
- AOI used to create spectral signatures (i.e., signature file)

Supervised Classification (Non-GIS)

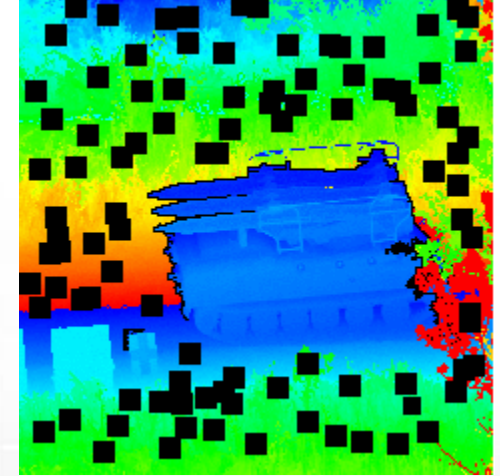
Sensor Image



All target pixels selected



Some clutter pixels selected



- Create ground truth (i.e., signature) file from selected pixels to train genetic algorithm to distinguish targets from clutter
- Each record (row) in file represents a pixel
- First field provides truth (1 = target; 0 = clutter)
- Subsequent fields provide features computed for pixel and neighboring cells



| Truth | x1 | x2 | x3 | |
|-------|-----|-----|-----|------|
| 0 | 84 | 20 | 156 | |
| 1 | 193 | 112 | 69 | |
| 0 | 91 | 28 | 134 | |
| 0 | 80 | 24 | 135 | |
| 1 | 193 | 102 | 79 | |
| : | : | : | : | |

Tools

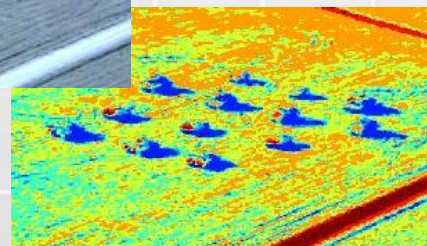
- Supervised Classification
- • Unsupervised Classification
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Unsupervised Classification

- Sorts raster pixels into individual classes (or, clusters)
- Algorithm determines which pixels are grouped together in classes; no need for *a priori* ground truthing
- But, analyst must decide what features those classes represent
- **GIS Tool**
 - ArcMap Spatial Analyst **ISO Cluster** and **Maximum Likelihood Classification**
 - ERDAS Image **Unsupervised Classification**



Original image



Classified image

Unsupervised Classification (GIS)

- Perform unsupervised classification to segment water feature from land features
- Study Area: Newnans Lake, Alachua County, FL

Landsat 5 Thematic Mapper Image
LT50170392011072GNC01

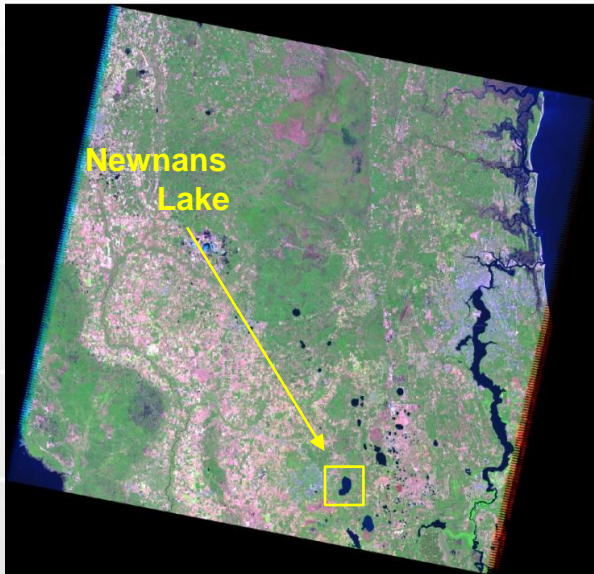
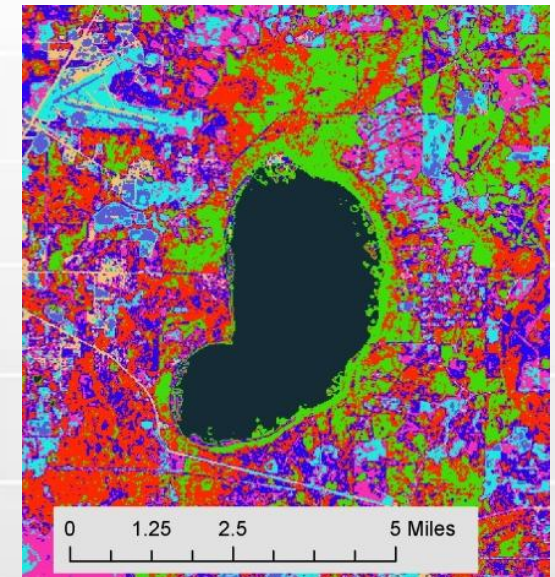


Image Chip



Classified Chip
(8 Classes)



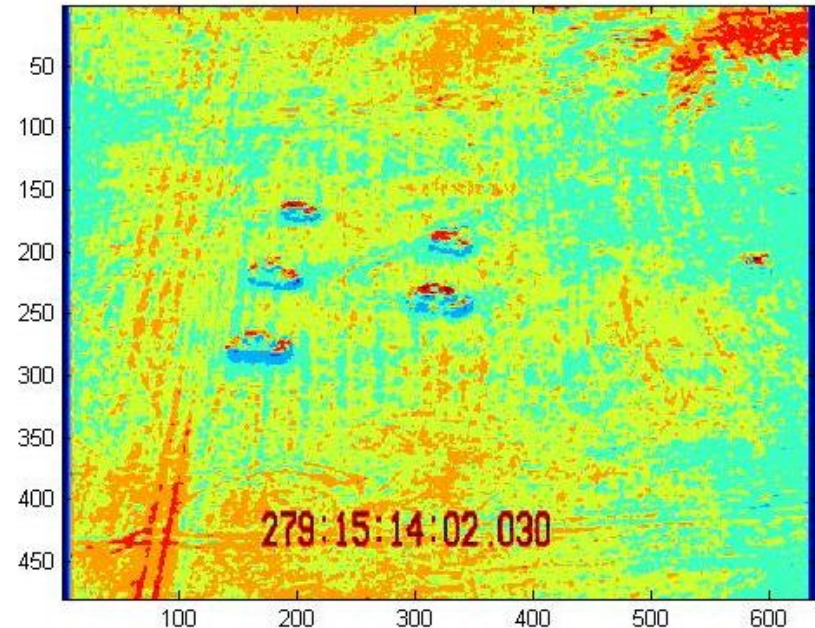
Unsupervised Classification (Non-GIS)

- Perform ***k-means clustering*** to segment image pixels into 'k' clusters to distinguish tanks from background clutter
- Pixels assigned to cluster with nearest mean value

Original Input Image



K-Means Clustering, k=8



Tools

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- Unsupervised Classification
- • Georeferencing
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Georeferencing

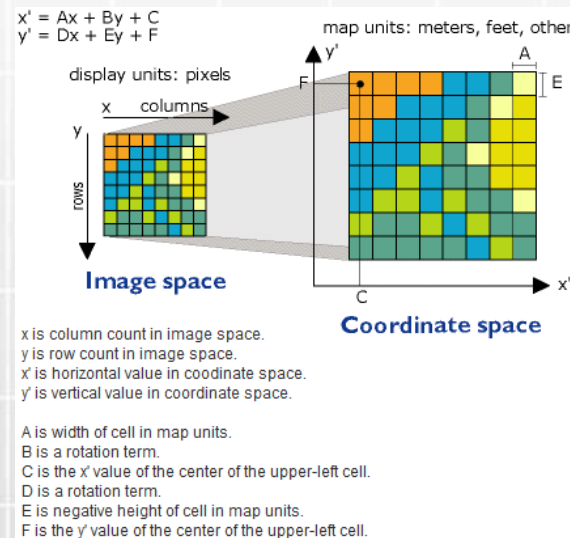
- Aligns raster/vector datasets to common coordinate system
- Control points used to link known positions in respective datasets
- One dataset is translated, rotated, and scaled relative to second dataset to achieved georeferenced image
- First-order polynomial (i.e., affine transformation) commonly used

- **GIS Tool**

- ArcGIS **Georeferencing Toolbar**

- Add Control Points

- View Link Table



Courtesy of ESRI



System Dynamics International

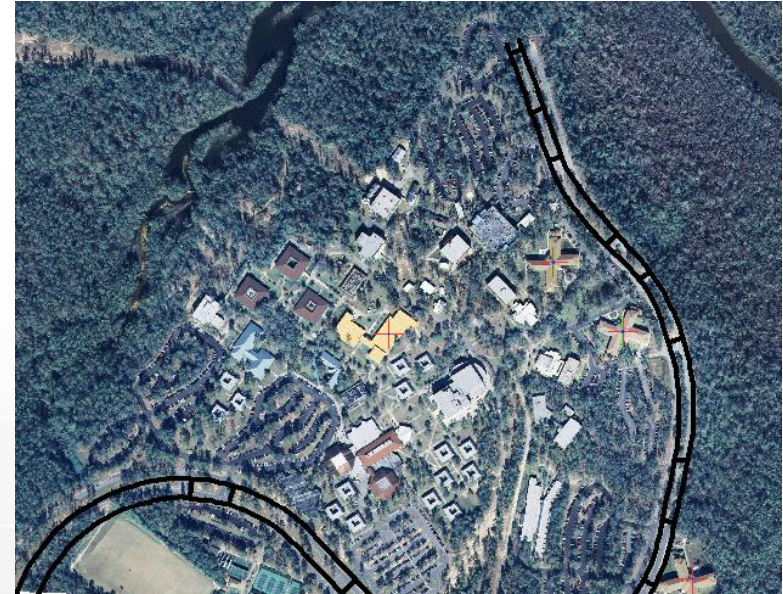


Georeferencing (GIS)

Vector data (roads and building) not georeferenced to aerial image raster



Georeferenced Image



Link Table

| Link | X Source | Y Source | X Map | Y Map | Residual |
|------|-------------|--------------|----------------|---------------|----------|
| 1 | 2786.522996 | -1351.042923 | 1113678.729996 | 574539.179993 | 3.43728 |
| 2 | 3142.826686 | -1685.098062 | 1114026.064588 | 574191.273097 | 3.68743 |
| 3 | 3551.505762 | -2913.858718 | 1114374.299997 | 572959.280008 | 0.96518 |
| 4 | 1981.656087 | -1737.864492 | 1112857.670657 | 574179.898257 | 0.71503 |

Transformation: 1st Order Polynomial (Affine)

Total RMS Error: 2.59109

Buttons: Load..., Save..., Restore From Dataset, OK

Link Table

rms = 2.59

1st-order polynomial
(affine transformation)

Georeferencing (Non-GIS)

Color
(Converted to grayscale)



IR



Color/IR Overlaid



- Georeferencing often referred to as **image registration**
- Although synchronized in time, vehicles in color image are offset (specifically, left and below) from corresponding vehicles in IR image
- Must “register” images to a single coordinate frame
 - Facilitates and optimizes data fusion from the two sensors

Image Registration (via MATLAB)



- Form following matrix equation, and solve for A (Affine Transformation matrix)

$$\begin{array}{c} \text{IR control points} \\ \begin{bmatrix} 293 & 175 & 240 \\ 185 & 235 & 355 \\ 1 & 1 & 1 \end{bmatrix} \end{array} = A \begin{array}{c} \text{Color control points} \\ \begin{bmatrix} 274 & 174 & 228 \\ 163 & 207 & 316 \\ 1 & 1 & 1 \end{bmatrix} \end{array}$$

Image Registration (Cont.)

$$\mathbf{A} = \begin{bmatrix} 293 & 175 & 240 \\ 185 & 235 & 355 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 274 & 174 & 228 \\ 163 & 207 & 316 \\ 1 & 1 & 1 \end{bmatrix}^{-1} = \begin{bmatrix} 1.18 & 0.009 & -33.05 \\ -0.128 & 1.107 & 8.02 \\ 0 & 0 & 1 \end{bmatrix}$$

Overlaid Registered Images



- Register color image to IR image by multiplying each color pixel $(\text{row}, \text{col}, 1)^T$ by transformation matrix, \mathbf{A}

Tools

- **Supervised Classification**
- **Unsupervised Classification**
- **Georeferencing**
- ➔ • **Spatial Filtering**
 - Low Pass (smoothing, averaging)
 - High Pass (sharpening)
 - Edge Detection
- **Lidar Point Cloud Processing**

Spatial Filtering

- Spatial filter (aka, kernel or mask) is group of cells around target cell
- Filter sequentially shifted across each cell in raster data set to recalculate value of target cell that lies at its center
- GIS Tool
 - Spatial Analyst **Filter-Low** for image smoothing
 - Spatial Analyst **Filter-High** for image sharpening
 - Spatial Analyst **Contour** for edge detection

| | | | | | | |
|----|----|-----------|----|----|----|----|
| 20 | 21 | 19 | 20 | 24 | 25 | 26 |
| 22 | 24 | 25 | 23 | 25 | 28 | 29 |
| 23 | 25 | 26 | 28 | 32 | 31 | 33 |
| 24 | 29 | 28 | 30 | 32 | 32 | 34 |
| 23 | 32 | 31 | 32 | 29 | 30 | 32 |
| 22 | 30 | 26 | 33 | 30 | 31 | 30 |
| 22 | 29 | 28 | 29 | 31 | 32 | 31 |

Raster data
(28 is target cell)

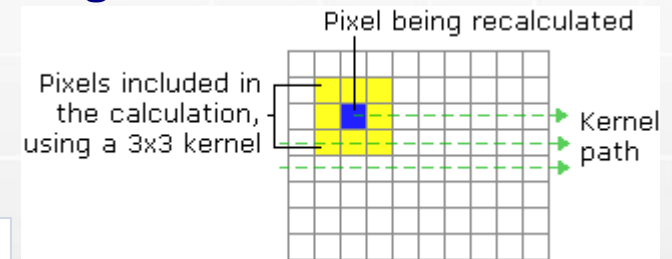
| | | |
|---|---|---|
| 1 | 1 | 1 |
| 1 | 1 | 1 |
| 1 | 1 | 1 |

Low Pass Filter

| | | |
|--|----|--|
| | | |
| | 29 | |
| | | |

Filtered value

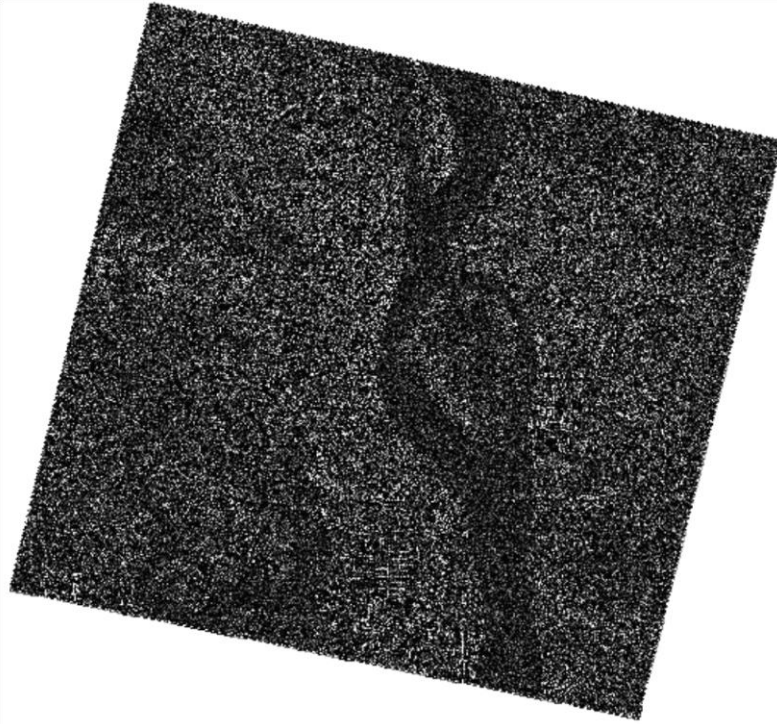
$$\begin{aligned}
 &(25 \times 1) / 9 + (26 \times 1) / 9 + (28 \times 1) / 9 + \\
 &(29 \times 1) / 9 + (28 \times 1) / 9 + (30 \times 1) / 9 + \\
 &(32 \times 1) / 9 + (31 \times 1) / 9 + (32 \times 1) / 9 = 29
 \end{aligned}$$



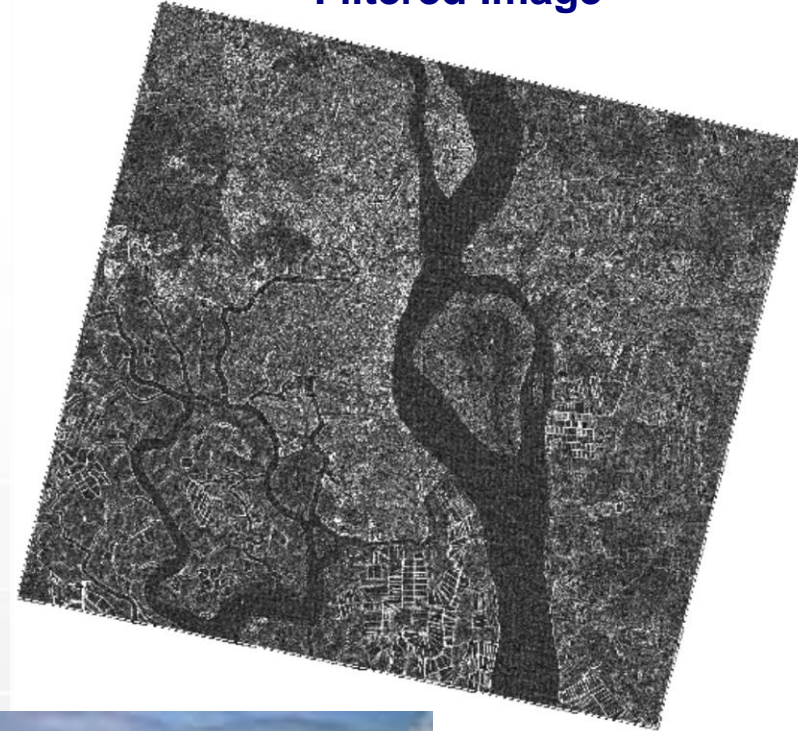
Courtesy of ESRI

Low Pass Filter (GIS)

Original Image (ETM, Band 8)



Filtered Image



- Filter smooths (averages) data
- Pixels blend better with neighboring pixels



Guayas River
Quayaquil, Ecuador

Low Pass Filter (Non-GIS)

Original Image



Filtered Image



- Filter smooths noisy image data

However ...

- Vehicle edges blurred and less distinct



High Pass Filter (Non-GIS)

Original Image



Filtered Image



- High-pass filter sharpens vehicle edges making them more distinct

However ...

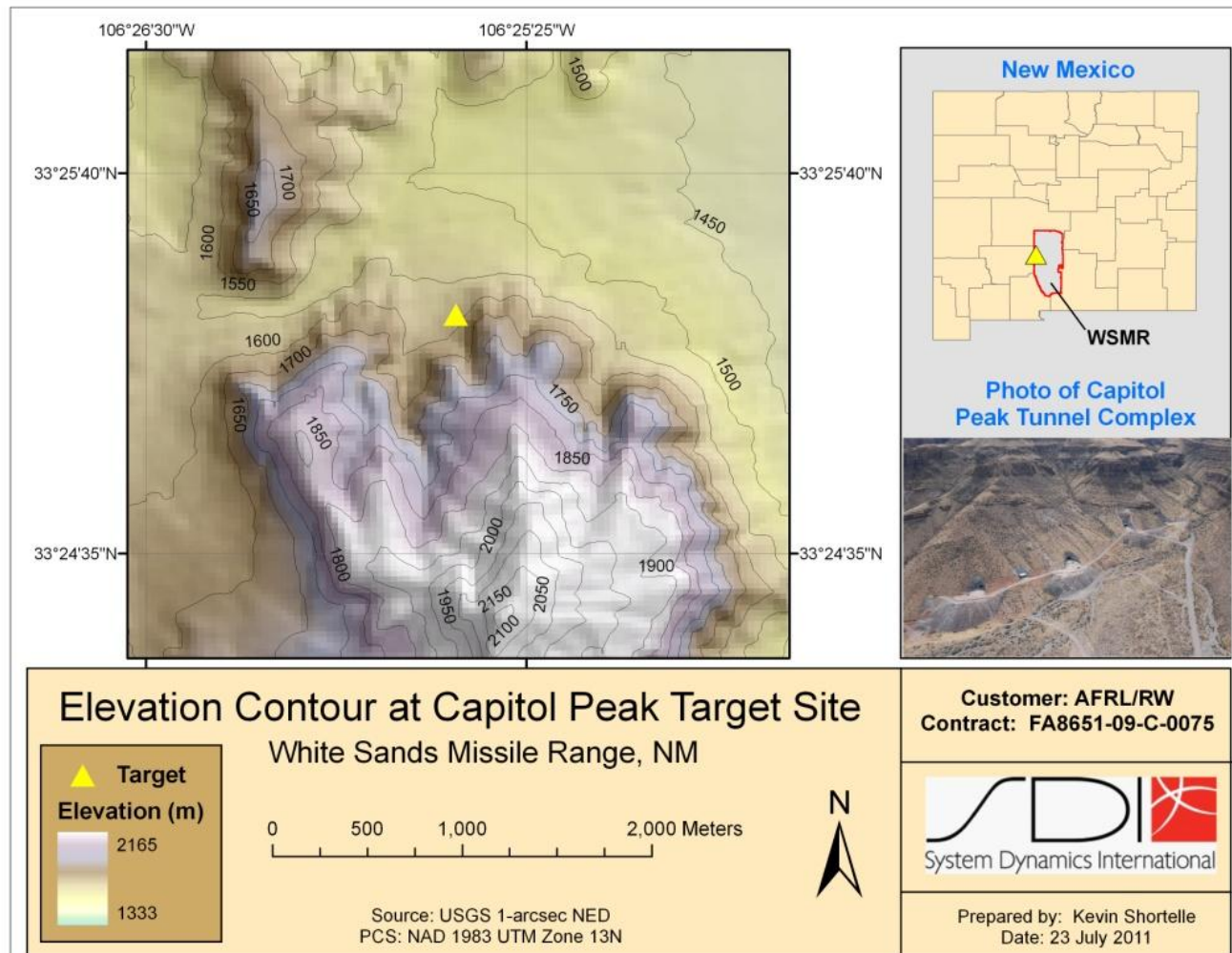
- Undesirable image noise also accentuated

| | | |
|----|----|----|
| 0 | -1 | 0 |
| -1 | 5 | -1 |
| 0 | -1 | 0 |

High Pass Filter

Edge Detection – Contour Tool (GIS)

- Identify gradients such as discontinuities and abrupt changes in DEM rasters to compute contour lines

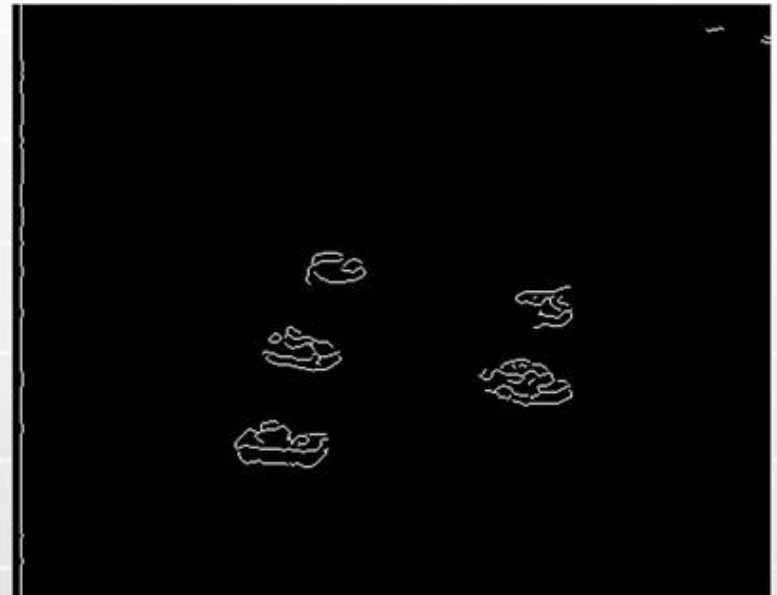


Edge Detection (Non-GIS)

- Filter used to identify distinct edges in raster image
- Result is reclassified two-code image -- value of 1 indicates a distinct edge, 0 otherwise
- "Sobel" filter uses a pair of 3 x 3 filters, one estimating the gradient in the x-direction (columns) and the other estimating the gradient in the y-direction (rows)

$$G_x = \begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix}$$

$$G_y = \begin{bmatrix} 1 & 2 & 1 \\ 0 & 0 & 0 \\ -1 & -2 & -1 \end{bmatrix}$$



Contour Tool Edge Detection (Non-GIS)

- As an experiment, use Spatial Analyst **Contour Tool** to segment military vehicles based on abrupt changes in grayscale values

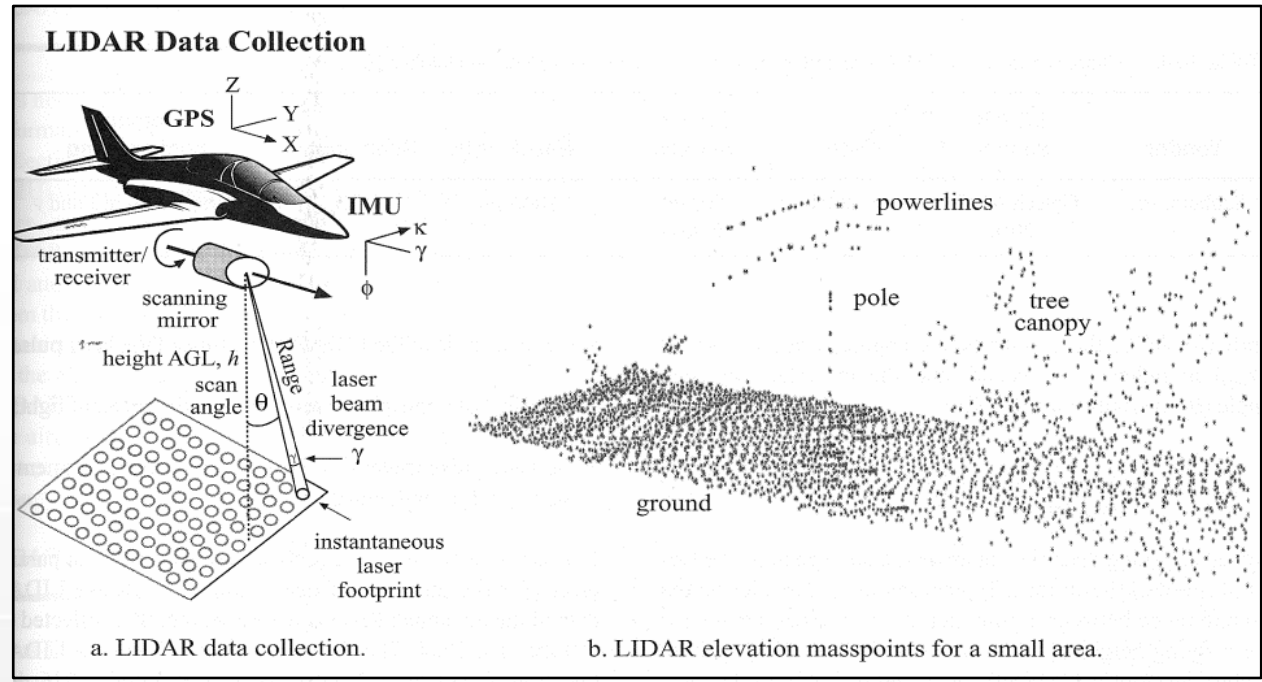


Tools

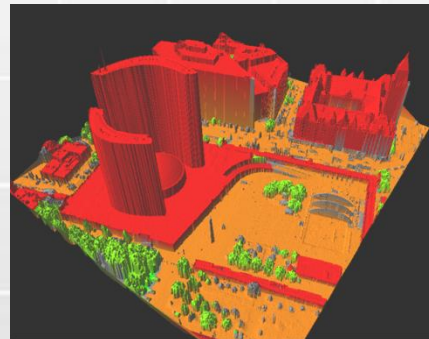
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- ➔ • **Lidar Point Cloud Processing**

Lidar Mass Point Processing

- 3D lidar data allows mapping of earth's surface and non-surface objects to render DTM and DSM
- Raw x, y, z data converted to LAS format to facilitate data exchange between users
- GIS Tools
 - ArcMap 3D Analyst LAS to Multipoint tool
 - Quick Terrain Modeler for point cloud visualization

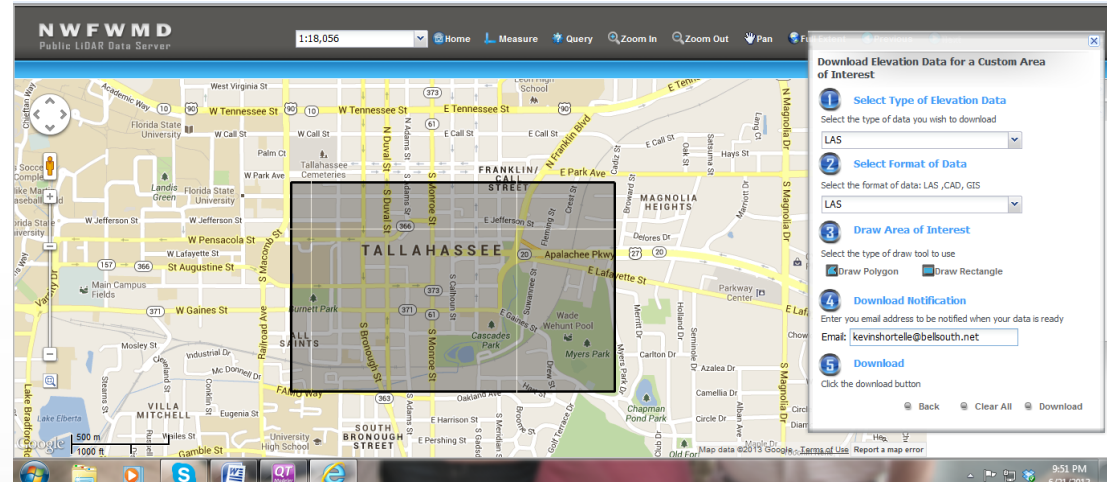


Text: Remote Sensing of the Environment

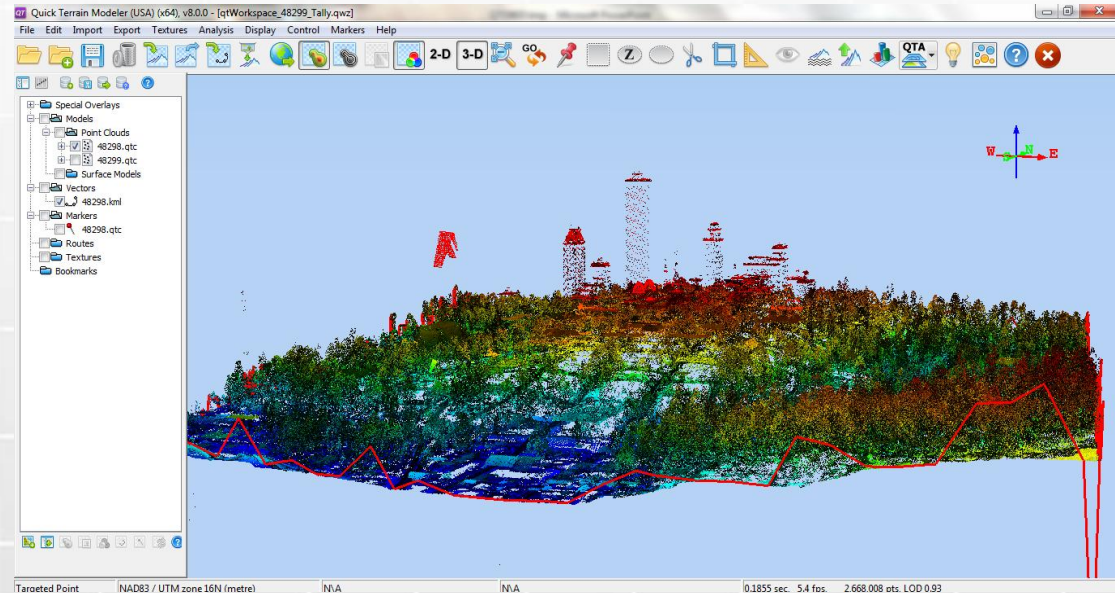


Lidar Mass Point Processing (GIS)

- Access NFWMD portal to retrieve Lidar data for selected area around Tallahassee, FL (www.nfwmdlidar.com)



- Utilize Quick Terrain (QT) Modeler to visualize retrieved data

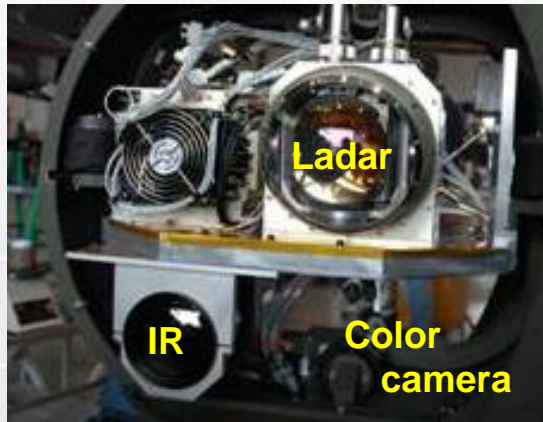


Ladar Point Cloud Processing (Non-GIS)

Exploit Ladar point cloud to develop automatic target recognition algorithms

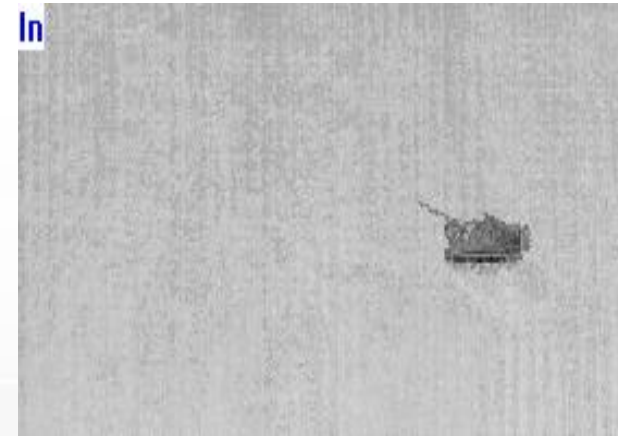


Stabilized Airborne Electro-optical Instrumentation Platform

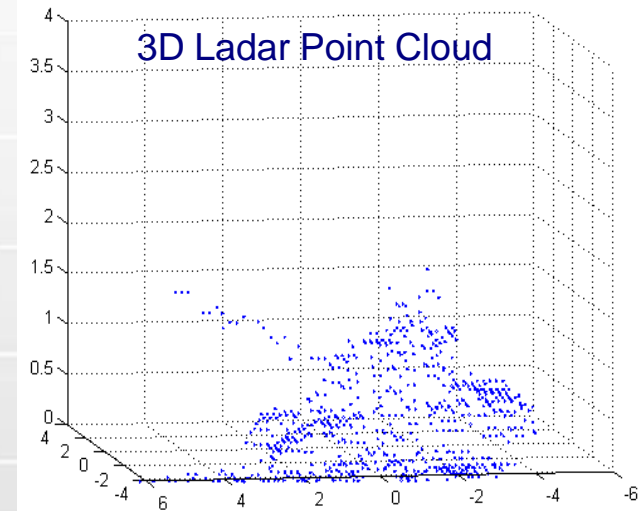


Co-located color, IR, and ladar sensors

2D Ladar Intensity Image



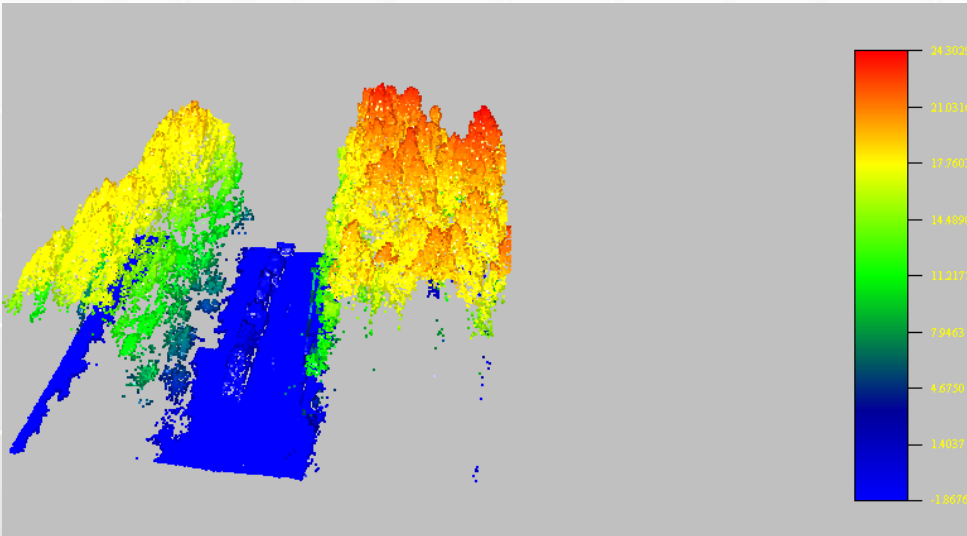
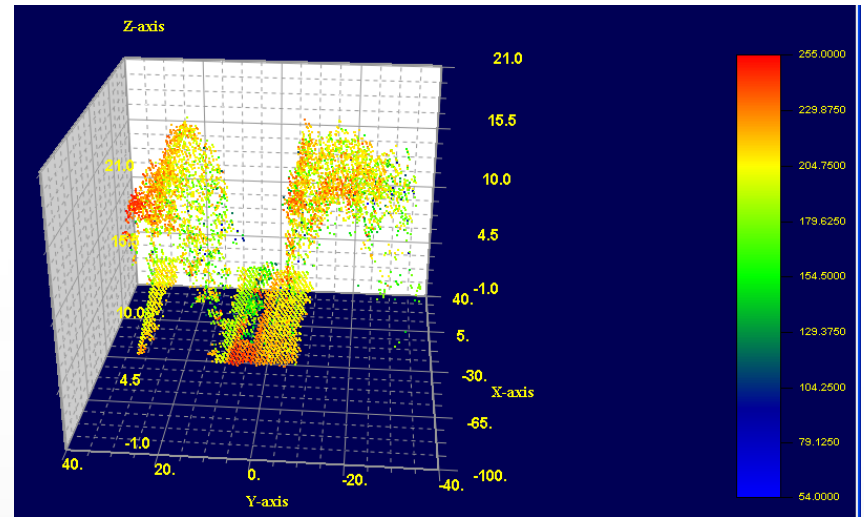
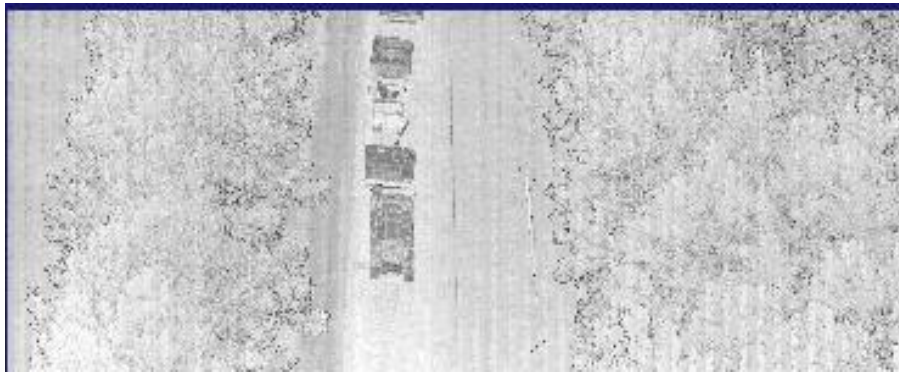
3D Ladar Point Cloud



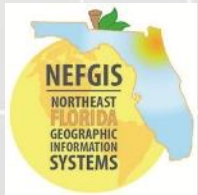
Ladar: Laser detection and ranging

Ladar Point Cloud (Cont.)

2D Ladar Intensity Image



Questions?



Supplemental Slides



Supervised Training Example (Non-GIS)

TRAINING DATA USED FOR TARGET DETECTION

| | Truth | FEATURE VALUES | | | | |
|---------|-------|----------------|-------|--------|-------|-------|
| | | x[1] | x[2] | x[3] | x[4] | x[5] |
| Target | 1 | 11.44 | 0.8 | 5.44 | 0.56 | 3.65 |
| | 1 | 5.6 | -0.29 | 3.53 | -1.07 | 2.83 |
| | 1 | 22.29 | 0.48 | 21.1 | -1.64 | 2.02 |
| | 1 | 23.72 | 0.33 | 26.75 | 6.35 | 8.46 |
| | 1 | 16.96 | 0.39 | 16.58 | -2.48 | 1.64 |
| | 1 | 129.88 | 0.95 | 146.47 | 15.08 | 7.05 |
| | 1 | 12.27 | 0.46 | 6.4 | 3.06 | 4.97 |
| | 1 | 11.89 | -0.15 | 7.46 | 8.4 | 2.34 |
| | 1 | 14.44 | 0.32 | 12.66 | -1.32 | -0.27 |
| | 1 | 21.12 | 0.29 | 22.49 | 3.06 | 7.69 |
| Clutter | 0 | 17.82 | -0.2 | 12.58 | -2.74 | -2.94 |
| | 0 | 9.06 | -0.87 | 3.84 | -2.83 | -1.88 |
| | 0 | 83.97 | 0.41 | 108.94 | 9.6 | 6.12 |
| | 0 | 138.36 | 0.6 | 108.94 | 6.43 | 1 |
| | 0 | 3.37 | -0.8 | 1.2 | -3.91 | -0.45 |
| | 0 | 49.98 | 0.35 | 108.94 | 5.83 | 7.72 |
| | 0 | 73.44 | 0.7 | 108.94 | 1.54 | 0.71 |
| | 0 | 7.13 | -1.2 | 3.18 | -4.75 | -1.3 |
| | 0 | 16.29 | -1.25 | 7.64 | -1.1 | -2.95 |
| | 0 | 17.49 | 0.05 | 17.78 | -4.67 | -2.53 |

RESULTING PSEUDO-CODE OUTPUT

```

Name of training data file = radar.dat
Hit miss info for best: hit/miss: (-)10/0 (+)10/0
[ 0] Registers initialized.

[ 1] Accum = Accum + X[5];
[ 2] Accum = Accum * X[1];
[ 3] Accum = Accum / X[3];
[ 4] Accum = Accum + X[5];
[ 5] Accum = Accum * X[4];
[ 6] Accum = Accum - X[3];
[ 7] Accum = Accum * X[4];
[ 8] if (Accum <= 0.0) Accum = 0.0; else Accum = 1.0;
    
```

$$\text{Accum} = \left[\left[\frac{x[1] * x[5]}{x[3]} + x[5] \right] * x[4] - x[3] \right] * x[4]$$

If Accum <= 0.0, Clutter; otherwise Target

K-means clustering

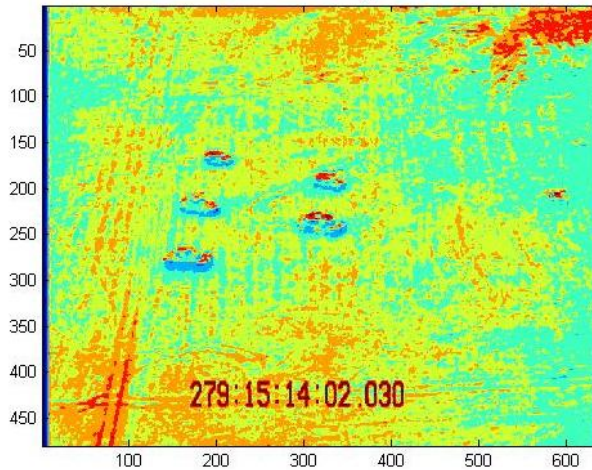
- Pixels assigned to cluster with nearest mean value

Cluster Means = [3.4037 60.4177 102.7171 134.4866 142.8289 151.2892 175.4506 233.2953]

Original Input Image

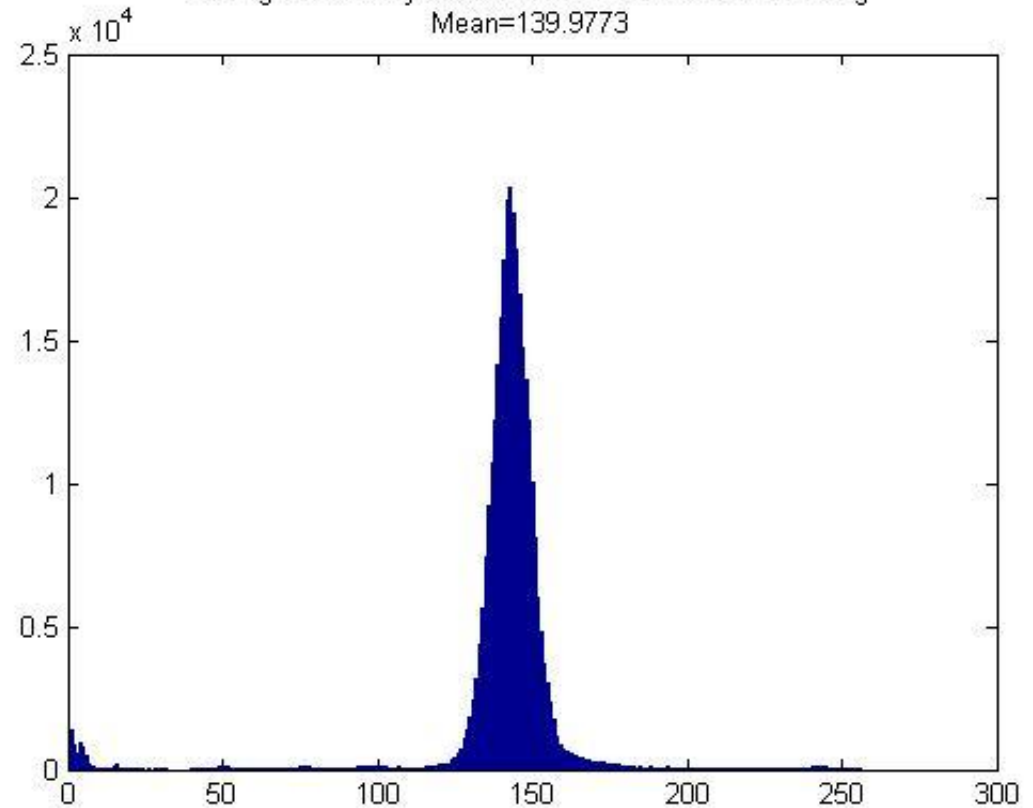


K-Means Clustering, k=8



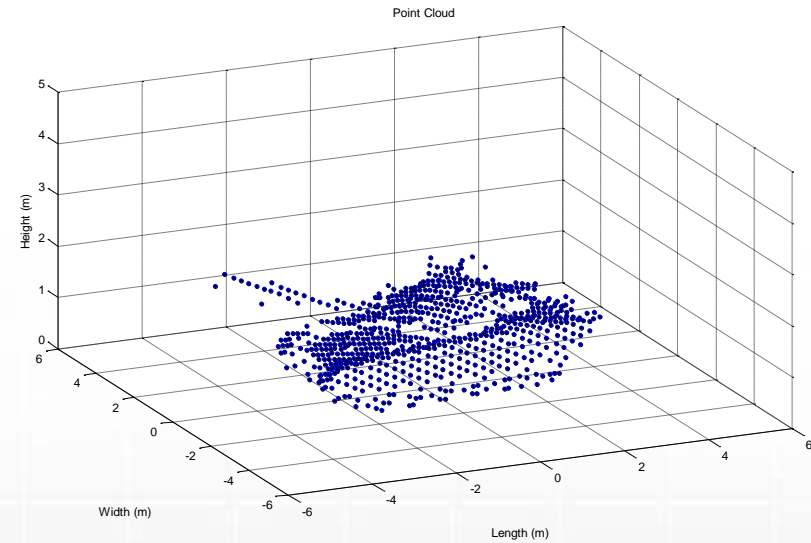
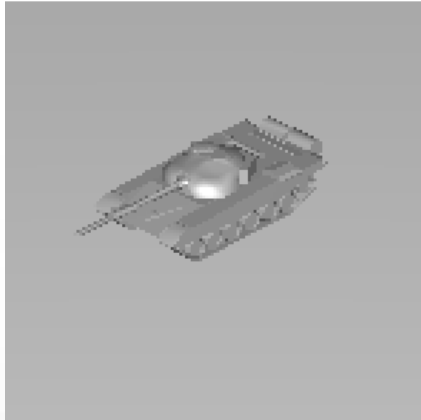
Histogram of Grayscale Values for K-means Clustering

Mean=139.9773



Ladar Point Cloud

Synthetically –Generated T72



- **MATLAB application provides interactive 3D rotation of rendered point cloud scene**

