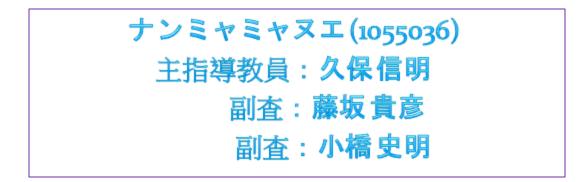
# Multi-Temporal Landsat Images Classification and Change Analysis of Land Cover in Urban Areas

Master's Thesis Defense 23 August 2012

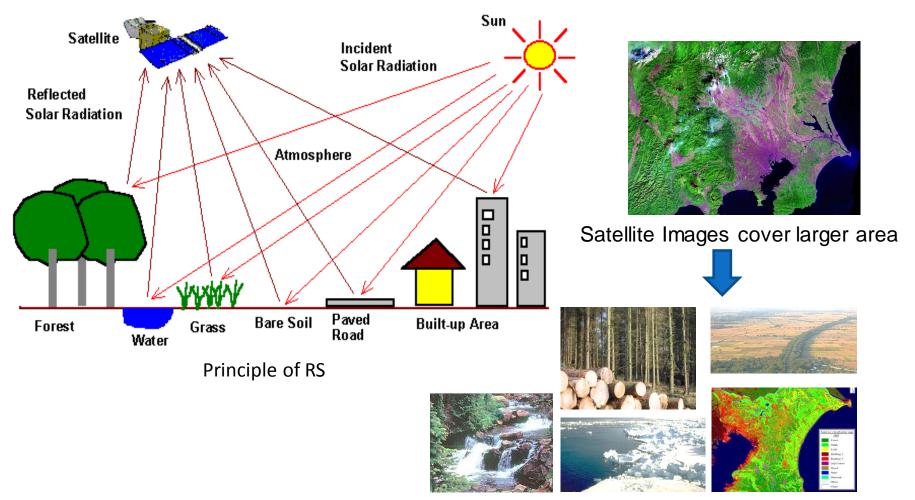


Tokyo University of Marine Science and Technology

#### Outline

- Introduction
- Aim and Purpose
- Study Area- Hino City
  - Data Acquisition
  - Image Classification
- Study Area-Mandalay city
- Evaluations & Conclusions

#### What is remote sensing?



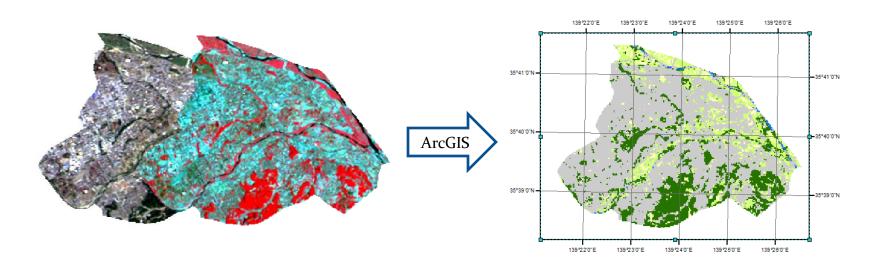
**Applications** 

### Introduction

- Land cover dynamics alter the availability of different biophysical resources \_ soil, vegetation, water and others.
- Satellite images -> accurate mapping & make landscape features and infrastructures .
- Geographic Information System (GIS) -> providing new tools for advanced ecosystem management.

# ArcGIS

- A geographic information system (GIS) for working with maps and geographic information.
  - Satellite Imagery >> improve and faster analyzing geographically information.

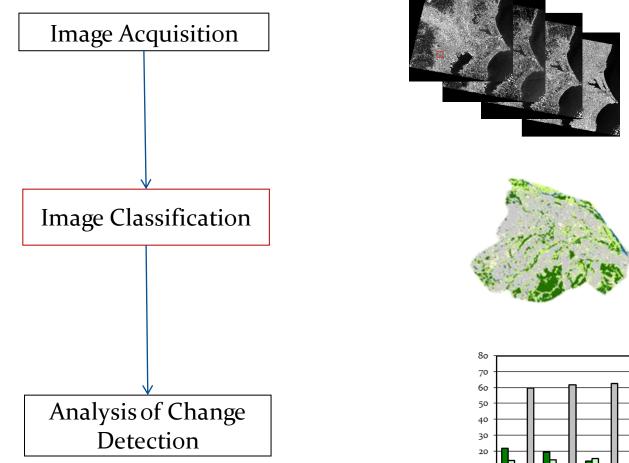


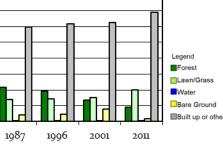
## Aim and Purpose

 to evaluate land cover & to analyze the change detection (RS Images analyzed by ArcGIS)

 intended to provide the accurate evaluation for managing natural resources and monitoring environmental changes.

# **Block Diagram of Study Analysis**





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# **Recent Studies on Land cover**

- Per-field classifier has shown to be effective for improving classification accuracy (Lloyd *et al. 2007*)
  - impact of the mixed pixel problem
- Neural network approach (Atkinson et al., 2009)
  - Not satisfied to shadows caused by topography
- Object-oriented classification approaches (Tanmoy Das, 2010)
  - still noticeable some misinterpreted classes

# **Classification approach**

- Multi-temporal images \*
- Colour composite \*
- Training Pixels by Knowledge based approach \*

# Study Area-1



Hino City Area= 27.53 km<sup>2</sup> Hino City in Western part of Tokyo Region

#### Multi-temporal Images

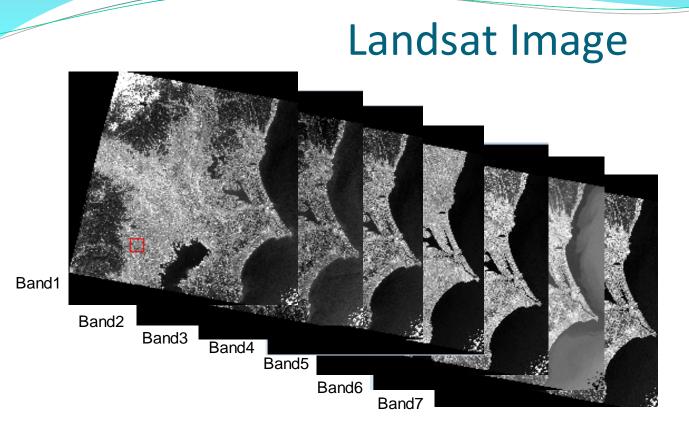
#### Data Source

No.	Data Type	Date of Production	Resolution	Source
1.	Landsat image (Hino city)	1987/5/21	30m™	USGS*
2.	Landsat image (Hino city)	1996/4/27	30m™	
3.	Landsat image (Hino city)	2001/4/1	30m <sup>⊾™+</sup>	
4.	Landsat image (Hino city)	2011/4/5	30m™	

5.	GIS Map of Hino City (IKONOS data-1m)	2008/9/9	1:56,424 (view scale)	Hino City
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\*United State Geological Survey TM- Thematic Mapper ETM-Enhanced Thematic Mapper

#### Knowledge for training pixel

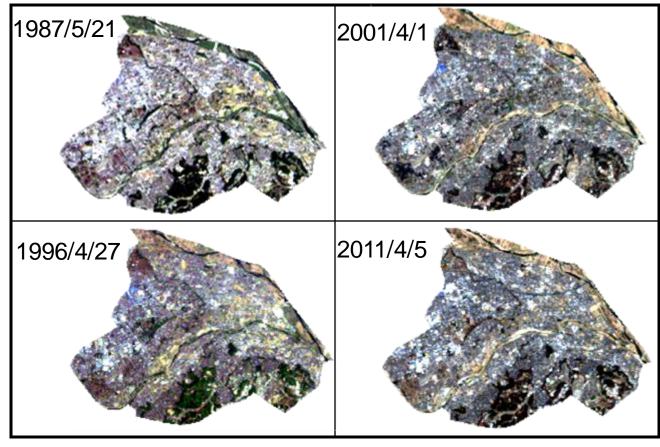


Pixel values – O(Black)~255(White)

7 bands of Landsat Image (Tokyo Region)

1)Blue (0.45-0.52μm)
2)Green (0.52-0.60μm)
3)Red (0.63-0.69μm)
4)Near Infrared (NIR) (0.76-0.90μm)
5)Intermediate Infrared (1.55-1.75μm)
6)Thermal (10.4-12.5μm)
7)Intermediate infrared (2.08-2.35μm)

#### Visual RGB Composite

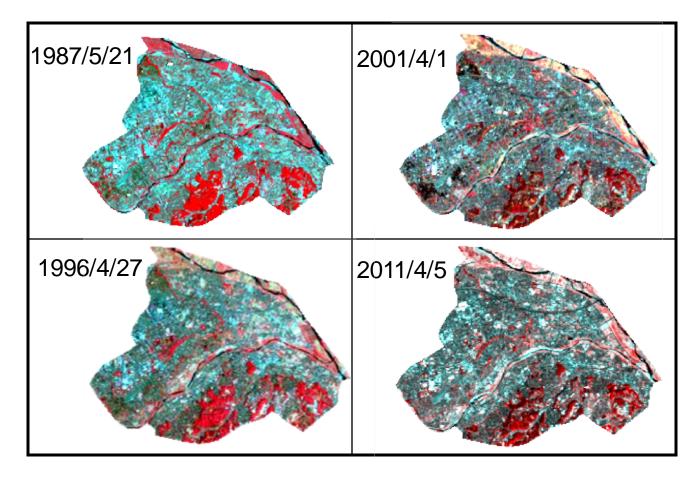


-Red band -> Red -Green band -> Green -Blue band -> Blue

- Used to classify type of ground

Landsat Images of Hino City

### **Near Infrared Composite**



-near Infrared Band -> Red

-Red band

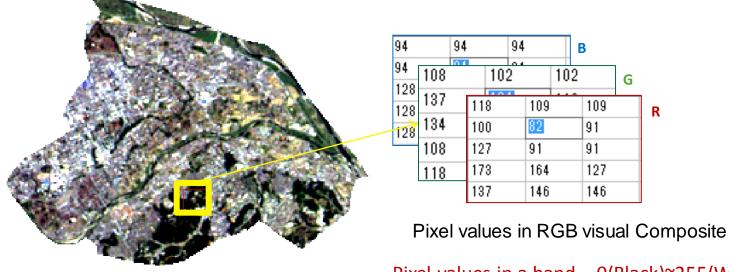
-> Green

-Green band -> Blue

- Used to classify vegetation and water.

Images of Hino City

#### **Pixels in Composite Image**

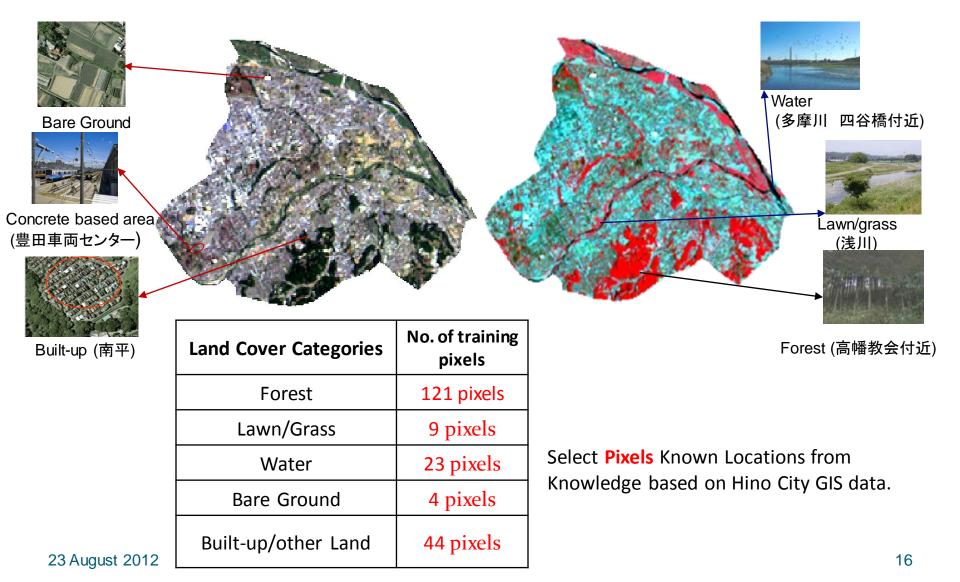


1987/5/21 RGB Visual Composite

Pixel values in a band – 0(Black)~255(White)

Total no. of Pixels in each band for Hino City – 30466 pixels

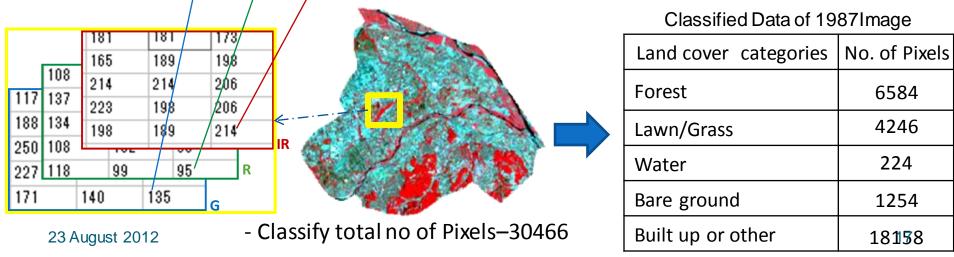
### Training Pixels by Knowledge based



# Classify the Image

Deciding the range of Training Pixels value

Land Cover	Mini ~ Maxi pixel value			Mean		
Land Cover	Green	Red	NIR	Green	Red	NIR
Forest	85~135	63~95	208~255	104.92	71.97	243.22
Lawn/Grass	127~170	109~137	150~222	155.88	127	197.25
Water	857120	68~105	10~59	99.95	81.09	29.45
	Blue	Green	Red	Blue	Green	Red
Bare Ground	29~32	30~35	32~44	30.56	32.88	38.31
Built-up/other	36~44	39752	36~56	40.15	46.08	49.10



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### **Classified Images**

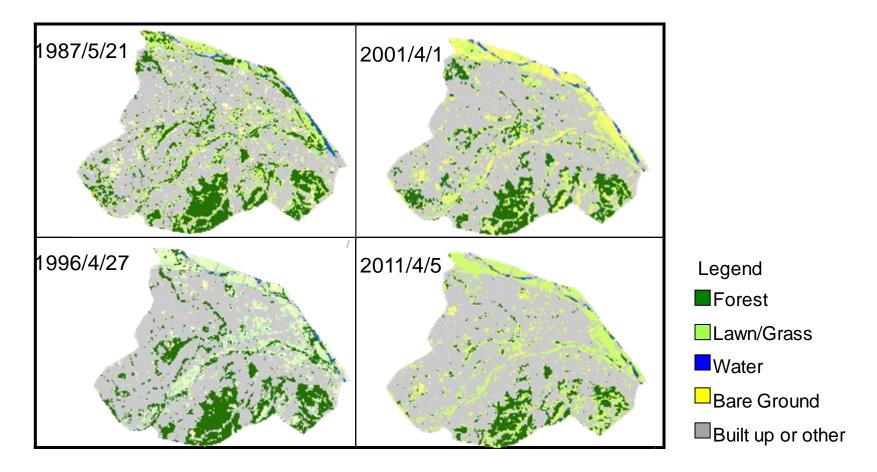
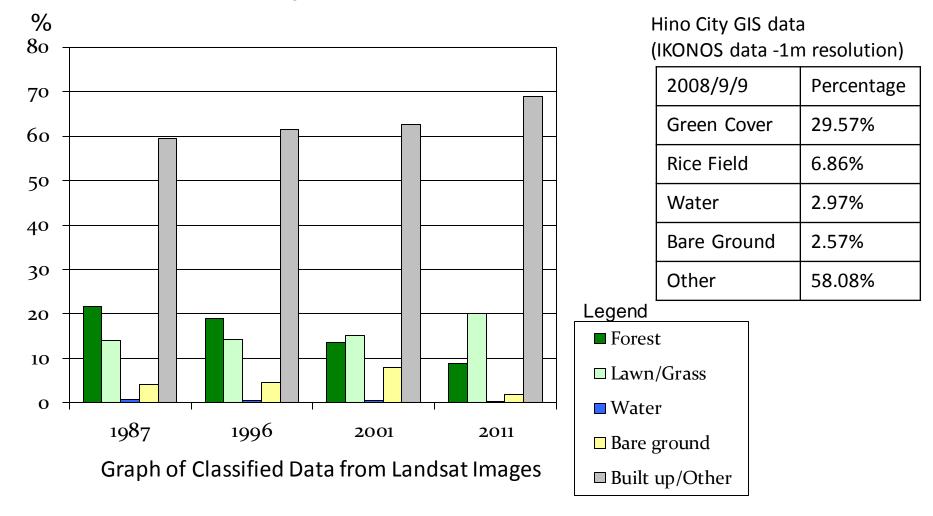


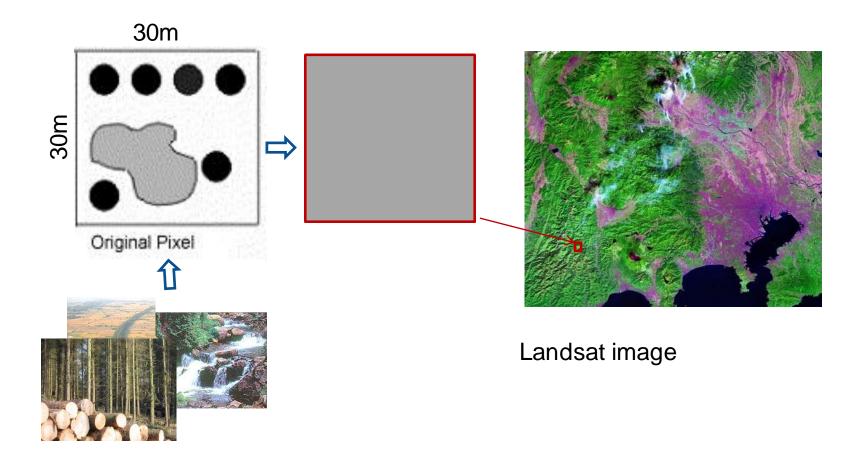
Figure 6. Classified Data of Hino City

#### Graph of Classified Data



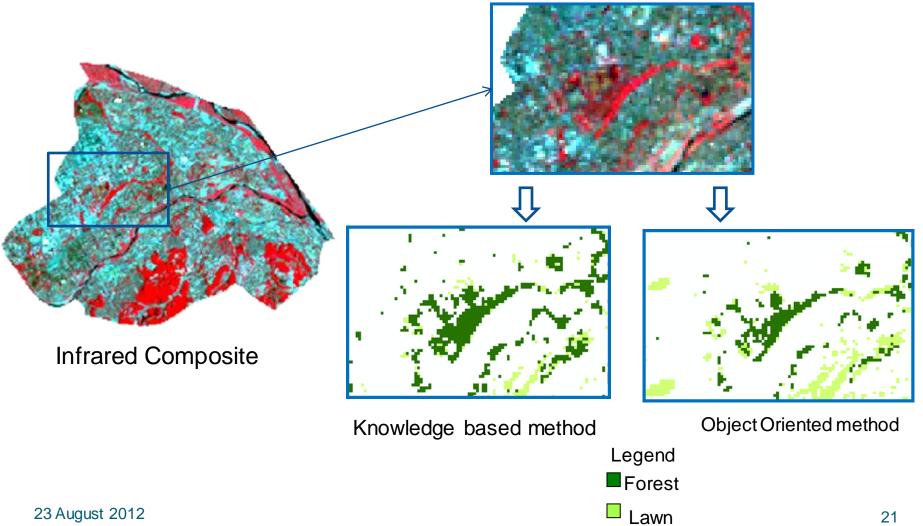
#### **Evaluations**

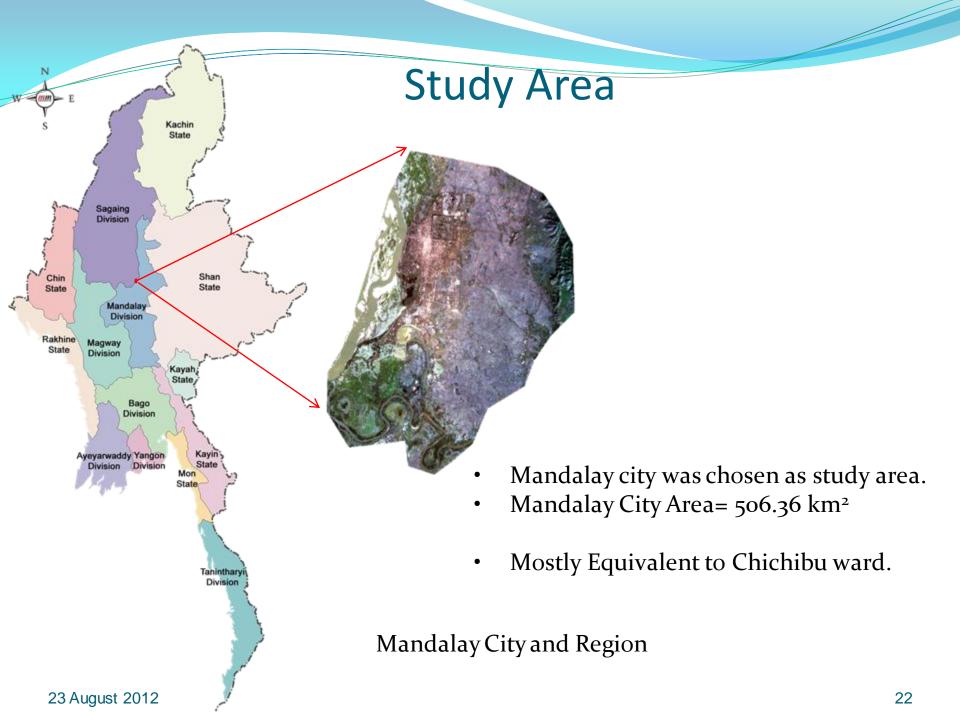
Mixed pixel problem can be ignored in Landsat image.



#### **Evaluations**

Every pixel was classified with no misinterpreted class.





# Multi-temporal Images

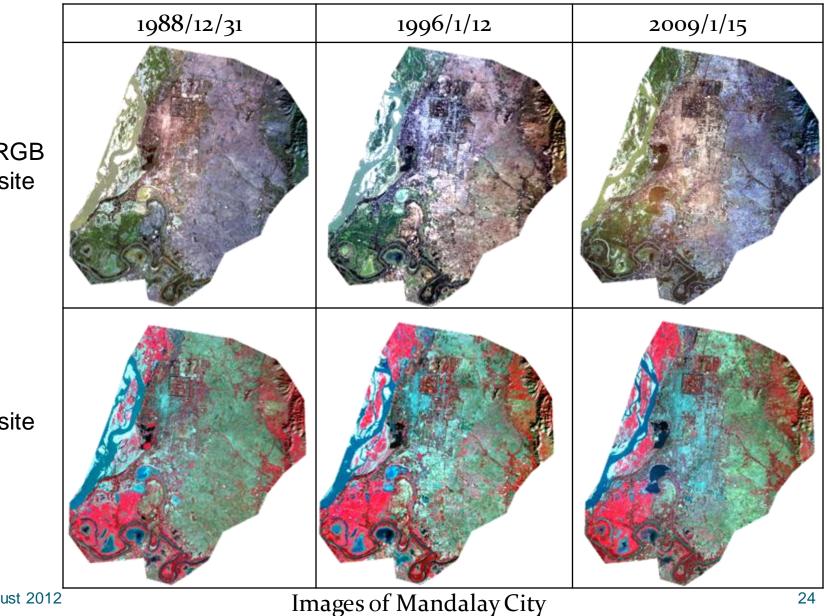
#### Data Source

No.	Data Type	Date of Production	Resolution	Source
1.	Landsat image	1988/12/31	30m™	USGS*
2.	Landsat image	1996/1/12	30m™	
3.	Landsat image	2009/1/15	30m™	

4.	Map of Mandalay City	2005	1:9,196,429 (view scale)	Local City Hall
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\*United State Geological Survey TM- Thematic Mapper

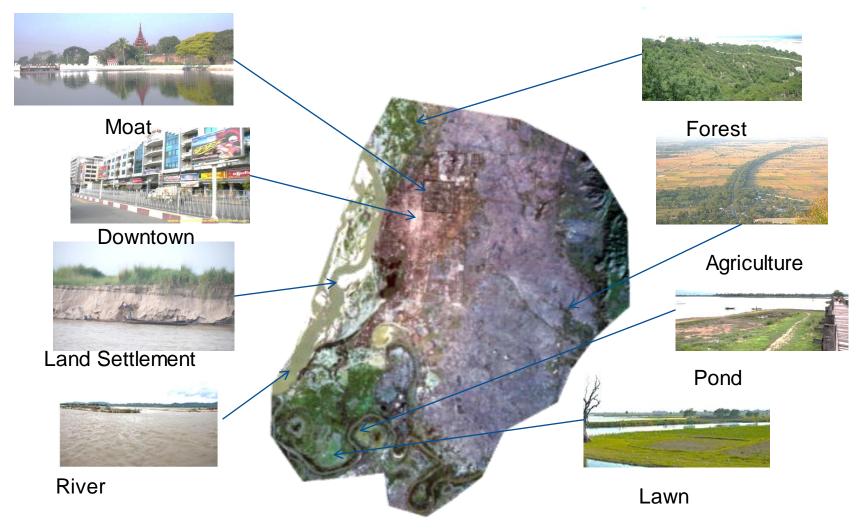
#### Visual RGB & Near Infrared Composite Images of Mandalay City



Visual RGB Composite

NIR Composite

### Training Pixels by Knowledge based



Known Locations from Map of Mandalay

# **Classify the Image**

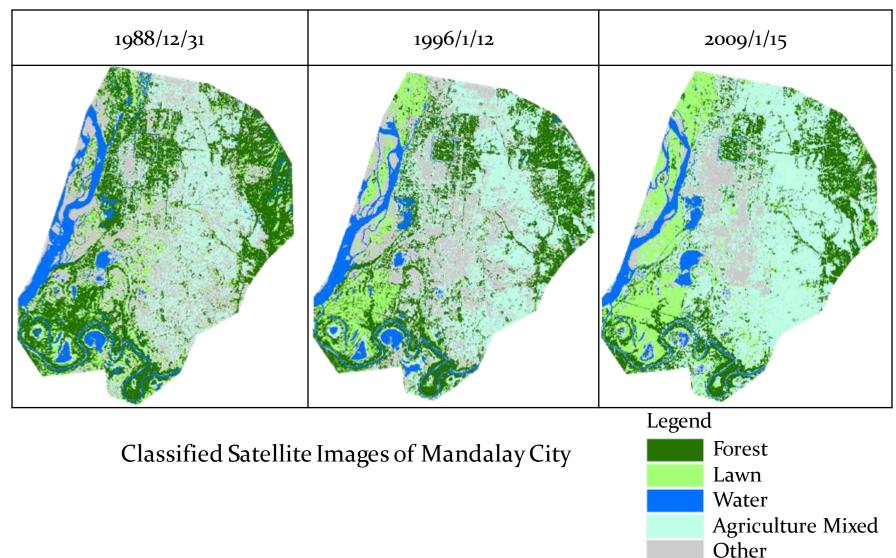
Deciding the range Training Pixels

	No. of	Mini~Maxi pixel value			Mean		
Land Cover	training pixel	Green	Red	NIR	Green	Red	NIR
Forest	199	23~28	19~26	46~113	26.34	21.52	64.47
Lawn/Grass	31	31~32	26~28	67~77	31.13	27.29	72.35
Water	1161	23~39	17~40	11~28	33.20	29.89	16.55
		Blue	Green	Red	Blue	Green	Red
Agriculture Mixed	16	29~32	30~35	<b>32~</b> 44	30.56	32.88	38.31
Built-up/other	84	36~44	39~52	36~56	40.15	46.08	49.10

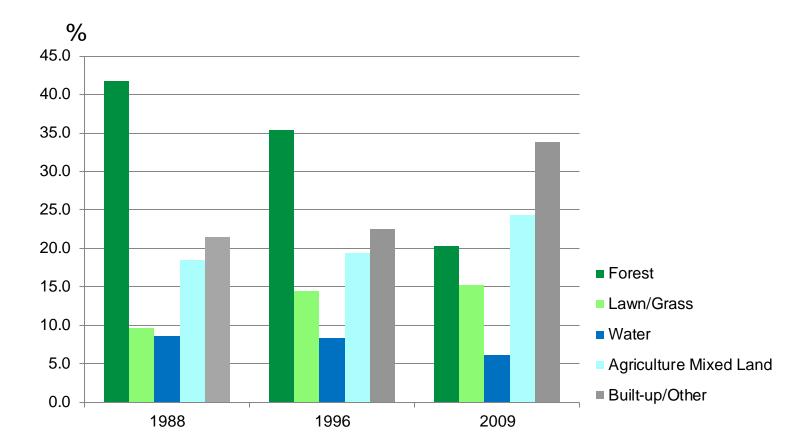
- Classify total no of Pixels–563929 (Image (1988) using ArcGIS)

Land cover categories	No. of Pixels
Forest	231211
Lawn/Grass	51754
Water	50753
Agriculture Mixed	107147
Built up or other	123064

### **Classified Images**



### Graph of Classified Data



Graph of Classified Data of Mandalay City

# **Evaluations**

- Works well on different landscape study areas
- Classified Five classes of land cover as through years.
  - Water Area changed due to land forming in Mandalay city.
  - Forest decreased greatly in both study areas.
  - Agriculture mixed land and other build up area increased in both study areas.

# Conclusions

- Demonstrates the ability of ArcGIS and RS
- Classification method is very suitable for both study areas using Landsat image. (Every pixel was classified to each corresponded class)
- The more training pixels, the higher accuracy.
- Depends on training pixels (not always a safe assumption from Map\*)
- Due to resolution of the images (30×30m)
  - not easy discriminating between light colored
  - Transitions between classes are always difficult
- To collect ground truth data by eco-social survey
- by utilizing very high resolution (VHR) satellite imagery

# **Publications**

- *"Correlation of Land Surface Temperature and Vegetation Density Classified from Satellite Images",* The proceeding of International Global Navigation Systems Satellite (IGNSS), Sydney, November (2011).
- *"Extracting the Assessment of Environmental Changes from Satellite Image",* The proceeding of International Symposium on GPS/GNSS, Tokyo, November (2008).
- *"Detecting the Environmental Changes from Satellite Image"*, The proceeding of 28<sup>th</sup> ASEAN Conference on Remote Sensing (ACRS 2007), Kuala Lumpur, November (2007), 202-203.

# Thank you very much for your attention!