

Urban Growth Analysis: Calculating Metrics to Quantify Urban Sprawl

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Project Overview

- ▶ Analyze change in urban spatial structure, over a 10 year period, for a global sample of 120 cities.
 - Phase I (Angel *et al.* 2005): Acquire and / or derive necessary data. Develop preliminary set of metrics
 - Phase II (Angel *et al.* 2007): Further develop metrics to quantify and characterize the spatial structure of the cities. Create maps to facilitate qualitative assessment of the urban structure

Phase I Study Objectives

- ▶ Select a stratified global sample of 120 cities with population over 100,000
- ▶ Classify urban land cover for each city at two time periods: circa 1990 (T_1) and 2000 (T_2)
- ▶ Derive preliminary metrics for characterizing urban development at T_1 and T_2

Phase I: Land Cover Derivation

- ▶ Land cover derived from Landsat satellite imagery.
- ▶ Land cover derived for two dates: T_1 (circa 1990) and T_2 (circa 2000).
- ▶ Land cover contained 3 categories: urban, water, and other

<u>Category</u>	<u>Grid cell value</u>
0	No Data
1	Other
2	Water
3	Urban

Classification of Built-Up Pixels

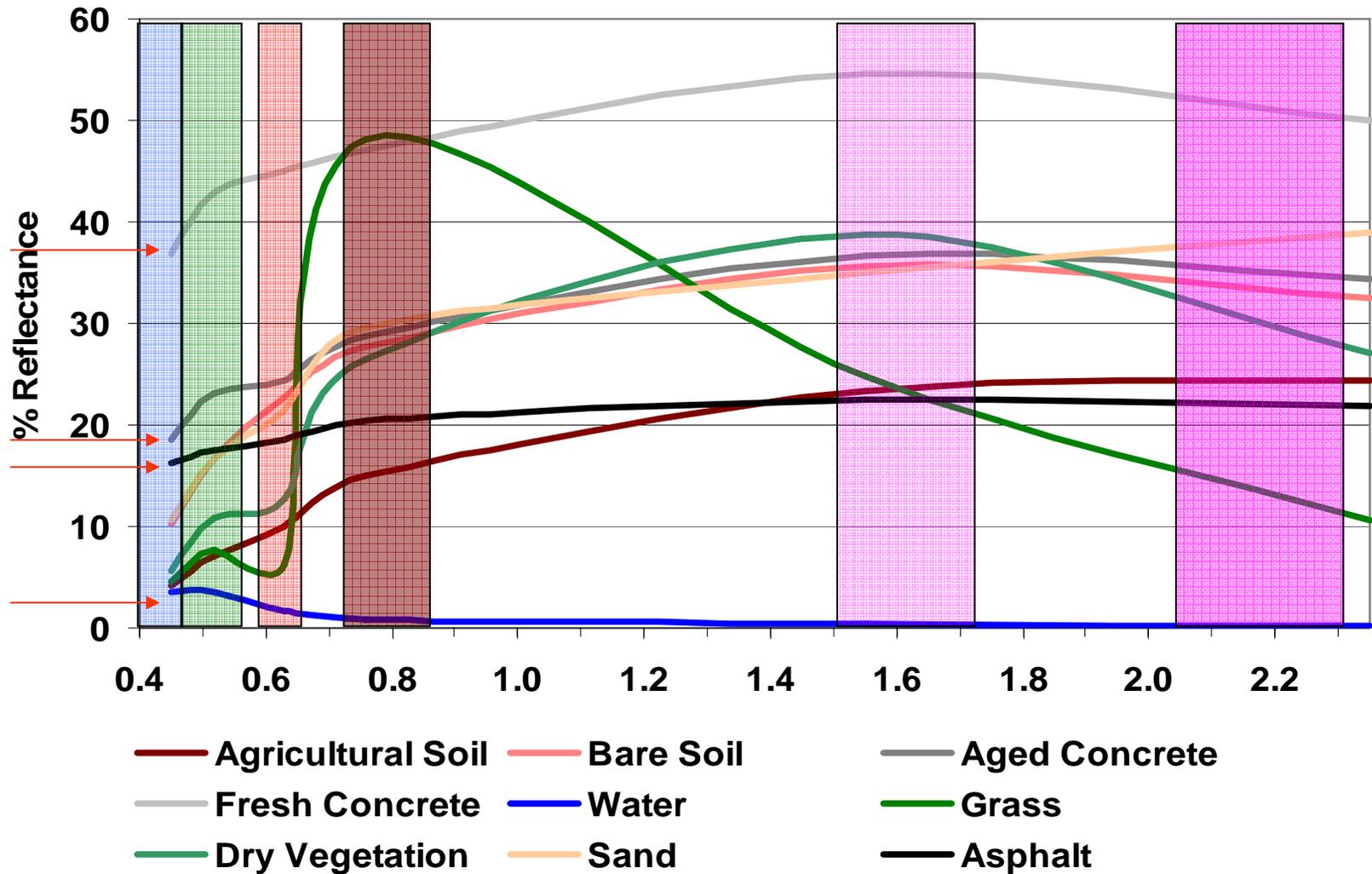
LANDSAT

- ▶ Satellite series operational since 1972
- ▶ Joint project between NASA and USGS
- ▶ Two sensors provided data for T_1 and T_2 dates: Thematic Mapper (LANDSAT 4 and 5) and Enhanced Thematic Mapper (LANDSAT 7)

TM and ETM Sensors

	Landsat 4/5 TM		Landsat 7 ETM	
Band	Spectral (μm)	Spatial (m)	Spectral (μm)	Spatial (m)
1 Blue	0.45-0.52	30	0.45-0.52	30
2 Green	0.52-0.60	30	0.52-0.60	30
3 Red	0.63-0.69	30	0.63-0.69	30
4 Near Infrared	0.76-0.90	30	0.76-0.90	30
5 Middle Infrared 1	1.55-1.75	30	1.55-1.75	30
6 Thermal Infrared	10.40-12.50	120	10.40-12.50	60
7 Middle Infrared 2	2.08-2.35	30	2.08-2.35	30
Panchromatic	n/a	n/a	0.50-0.90	15

Spectral Reflectance Curves



Classification of Built-Up Pixels

Image Acquisition and Processing

- ▶ Landsat scenes acquired for each study city.
 - As close as possible to 1990 or 2000
 - Minimal cloud cover
- ▶ Software: Leica Geosystem's Erdas Imagine 8.7
- ▶ Scenes clipped using preliminary set of administrative district boundaries

Classification of Built-Up Pixels

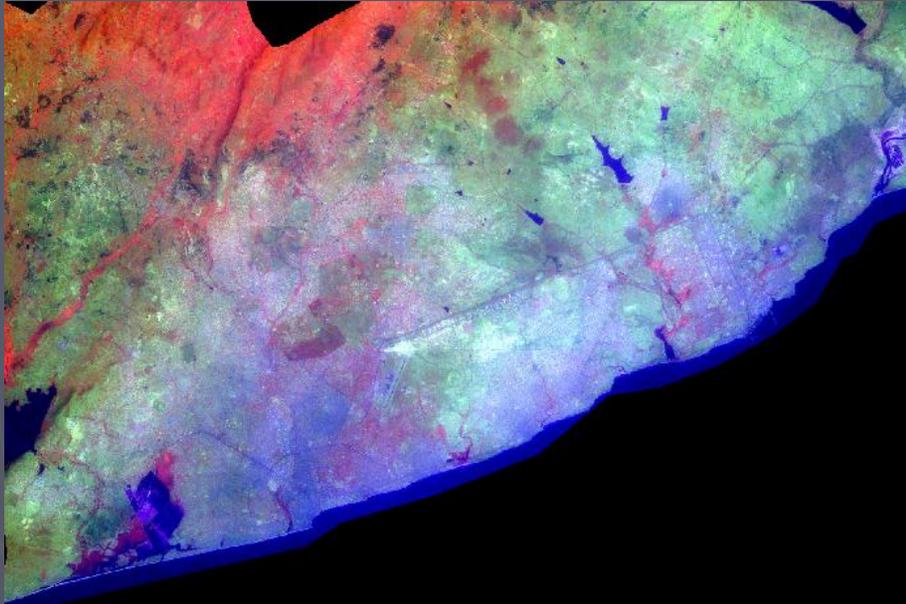
Unsupervised Classification method:

- ISODATA (Iteratively Self-Organized Data Analysis)
- Data assigned to homogenous classes based on spectral properties
 - ▶ Ideally classes correspond to a single land cover class
 - Certain types of land cover may have similar spectral properties
 - Shadows and clouds are problems

Classification of Built-Up Pixels

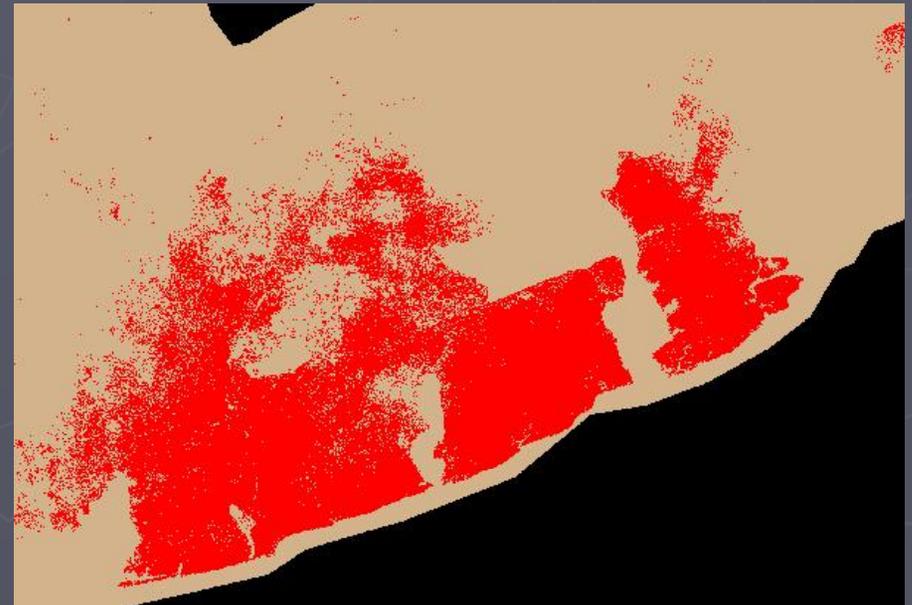
- ▶ User identifies land cover type that best corresponds to each ISODATA class
 - Occasionally multiple land cover types correspond to a single class (i.e. urban and barren land)
- ▶ Manually edit classification errors using on screen digitizing and recoding
- ▶ Recode edited classification image to two classes: urban and non-urban

Classification of Built-Up Pixels



LANDSAT for Accra T_2
(NIR, MIR, and green)

Unsupervised classification
and correction

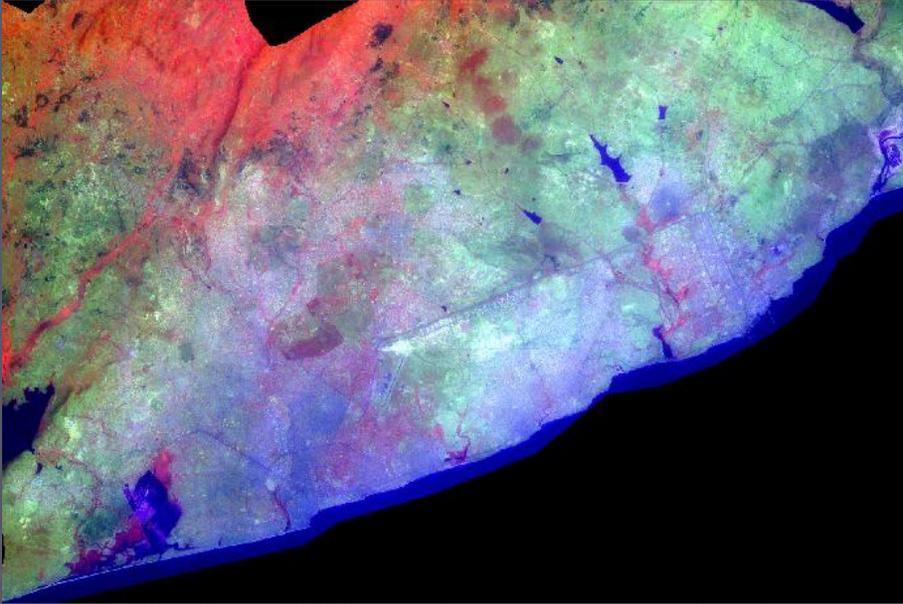


Urban / Non-urban Image 11

Classification of Water Pixels: The Water Index

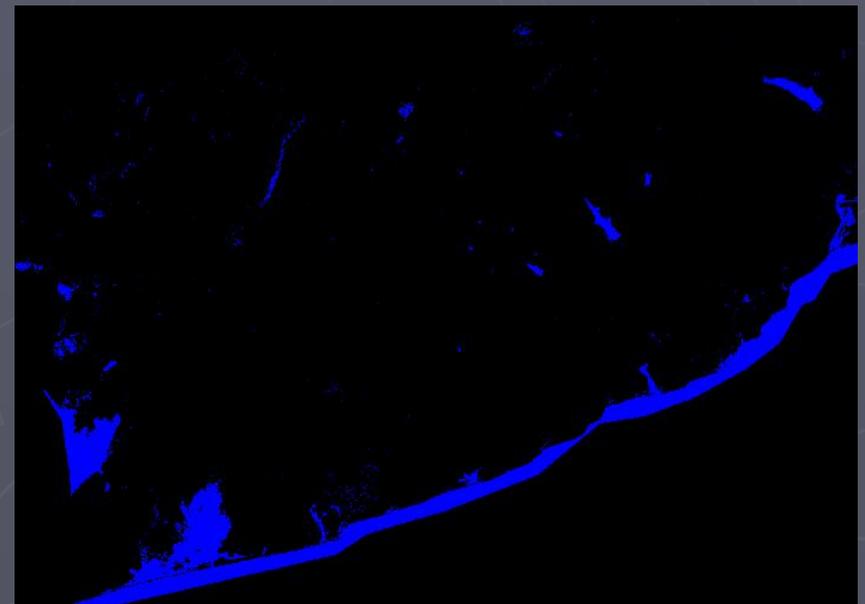
- ▶ Water has moderately low reflectance in visible bands (red, green, blue) and very low reflectance in infrared bands.
- ▶ Water index is the following ratio (scaled from 0-255):
 - Sum of visible bands / sum of infrared bands
- ▶ High water index values = water
 - Analyst identifies a threshold above which indices correspond to water.
- ▶ Recode water index image into two classes: water and not water.

Classification of Built-Up Pixels



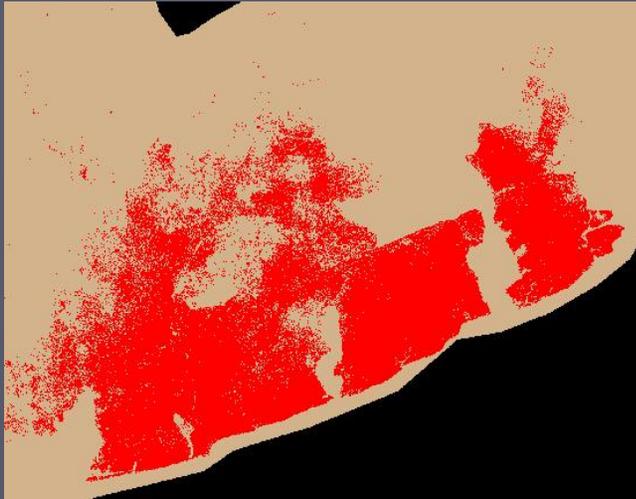
LANDSAT for Accra T_2
(NIR, MIR, and green)

Classification based on
water indices

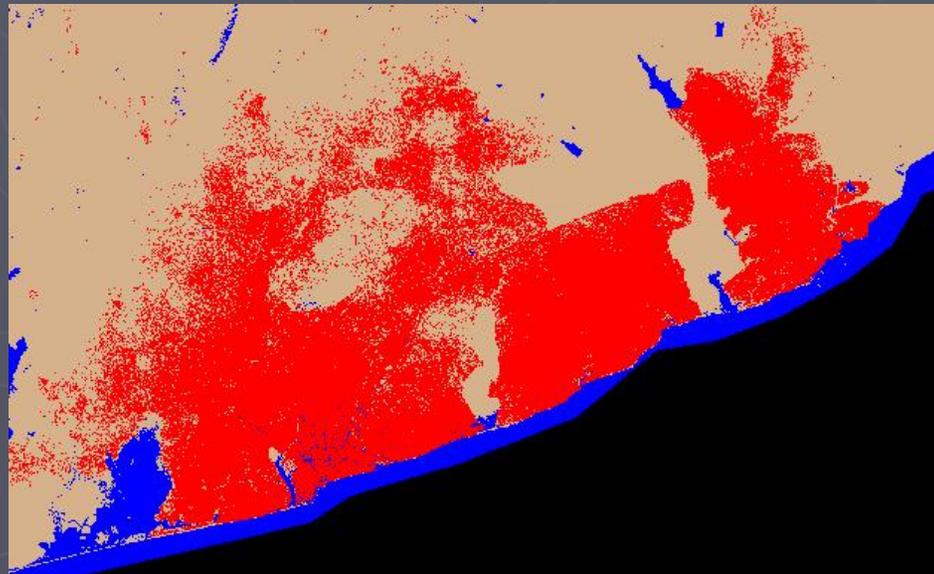
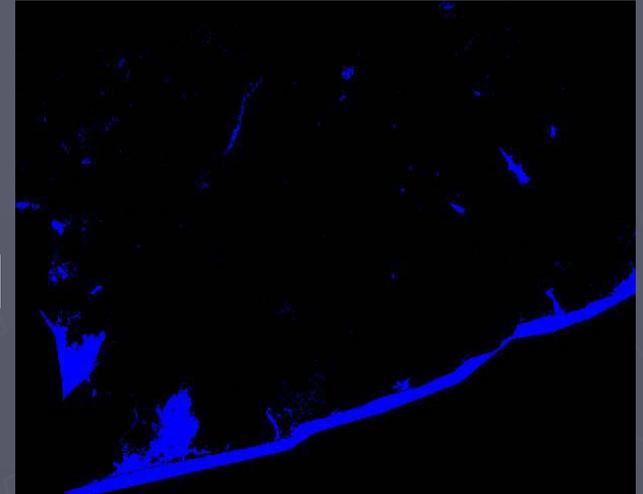


Water for Accra T_2 13

Combining the Urban and Water Land Cover Maps

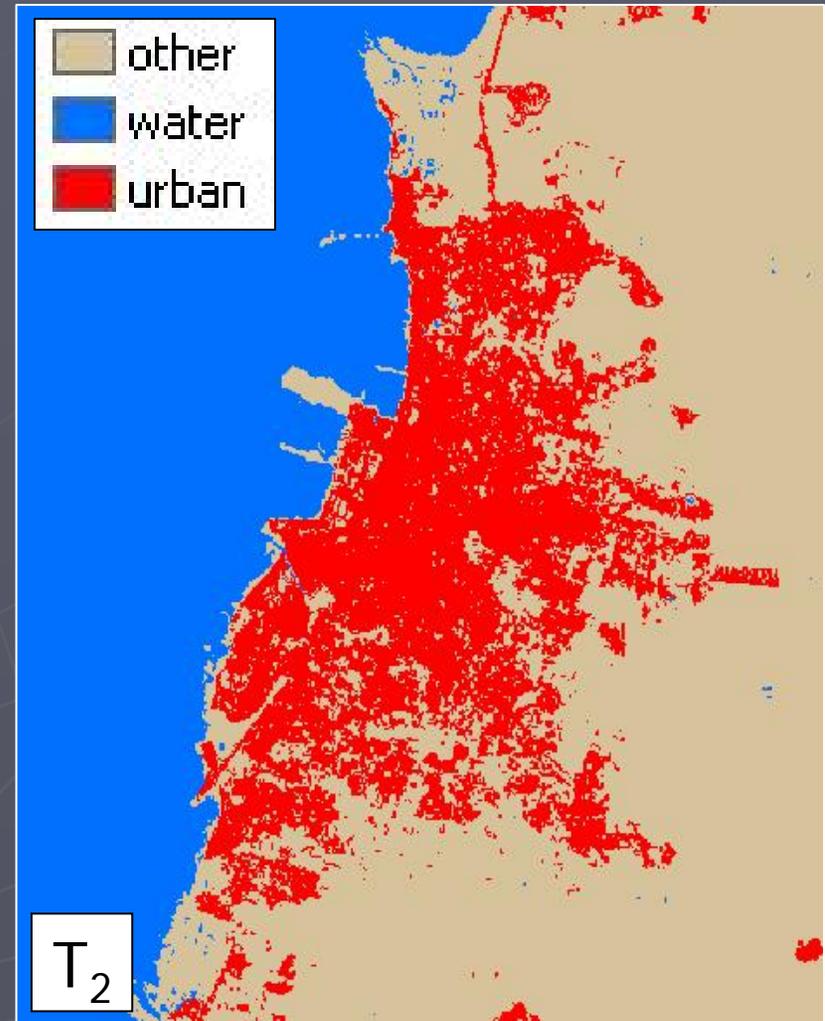
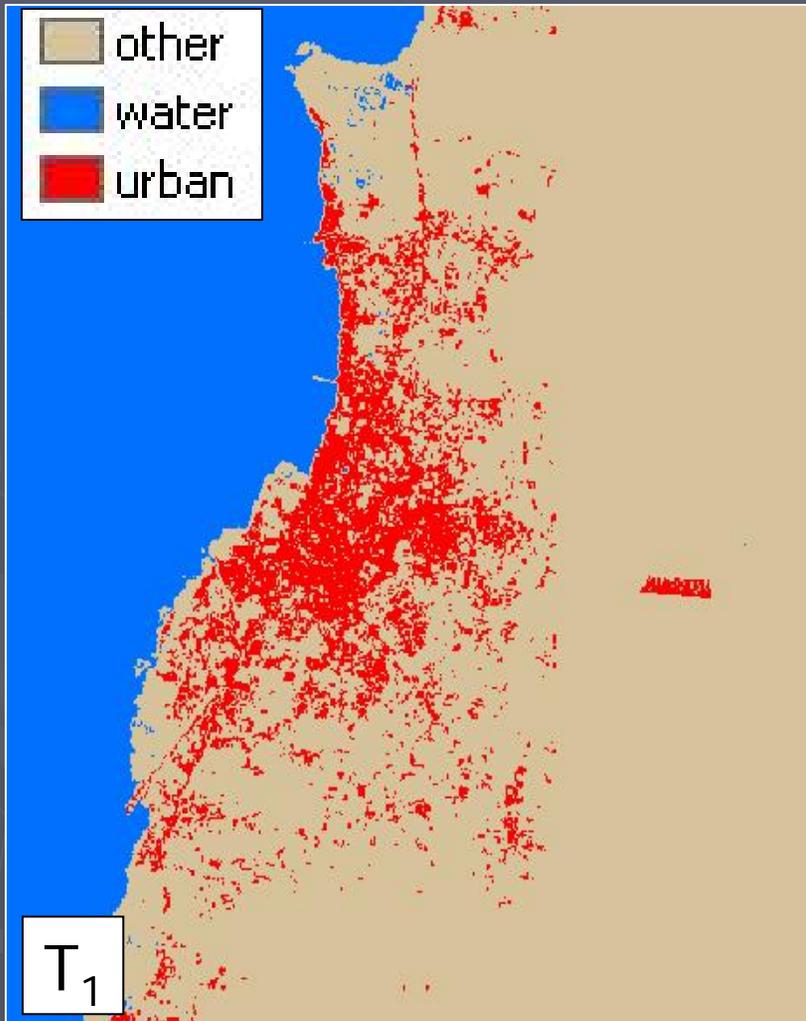


Overlay



Urban and Water Image for Accra T₂

Bacolod, Philippines: Land cover



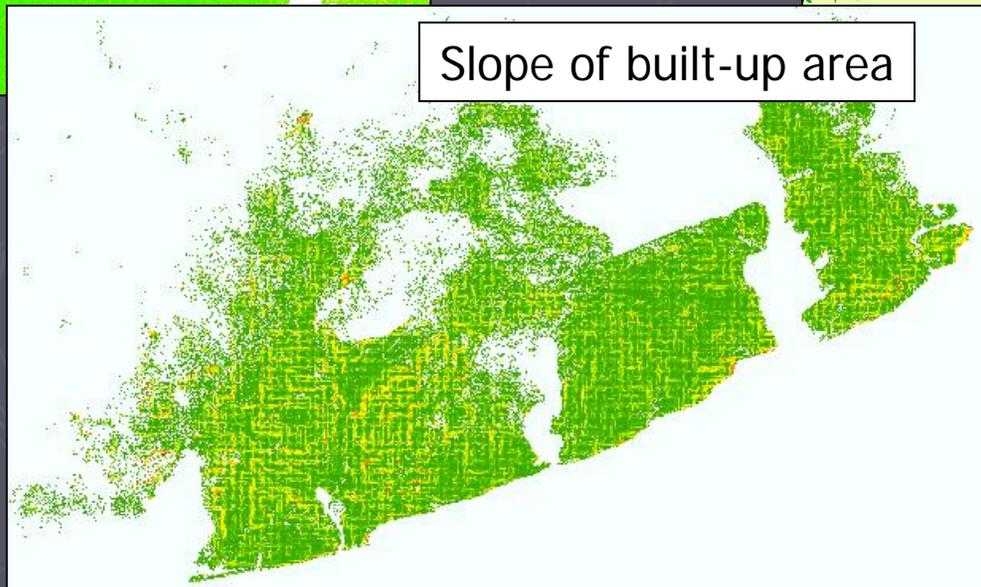
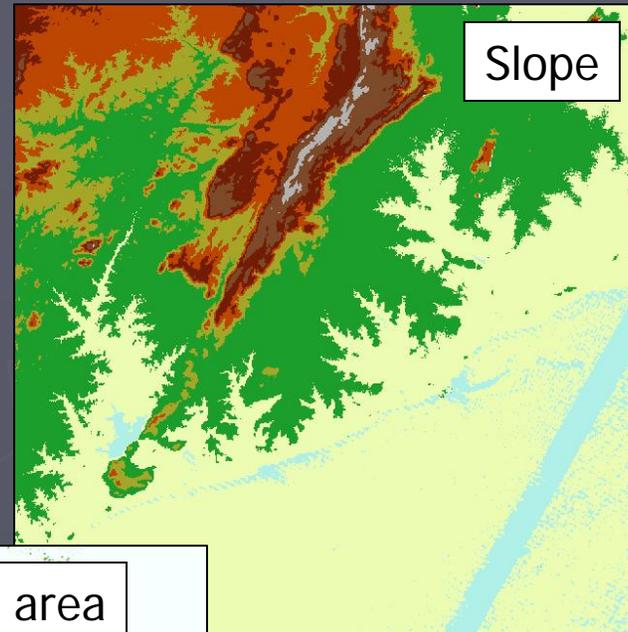
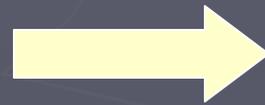
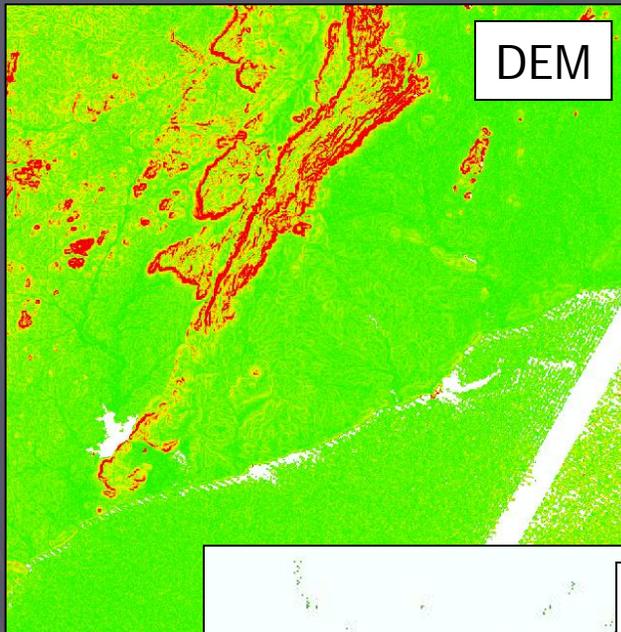
Phase I: Acquisition of City Boundaries

- ▶ Administrative districts (*i.e.* Census tracts) acquired for each city.
- ▶ Population data were attributed to the administrative districts.
 - Interpolated census data to estimate populations at time of land cover dates.
- ▶ Districts define the outer perimeter of the cities and provide population data.

Phase I: Slope

- ▶ Slope grid derived from Shuttle Radar Topography Mission (SRTM) data.
- ▶ Slope calculated in percent.
- ▶ Maximum slope defined as the slope value below which 99% of the urban area exists.

Determining Maximum Slope



Average and Maximum Slope

ObjectID	Value	Count
0	0	106772
1	0.620269119739532	7449
2	0.877192974090576	5391
3	1.24053823947906	7876
4	1.38696384429932	12689
5	1.75438594818115	52398
6	1.86080729961395	4942
7	1.96146309375763	1524
8	2.23641204833984	4674
9	2.48107647895813	5140
10	2.55743503570557	4108



Average and
maximum
slope



Slope



Pixel count

The maximum slope is the slope value at the 99th percentile

Phase II: The Metrics

- ▶ Several sets of metrics were developed to measure specific aspects of the urban spatial structure
- ▶ This presentation presents a sample of metrics from each set
 - Angel et. al. (2007)
 - The complete set of metrics will be presented in future papers at the conclusion of the study
- ▶ We present data for Bangkok and Minneapolis for illustrative purposes. Data for the analysis of the 120 cities will be presented in a future paper.

Sprawl Manifestations

- ▶ At a given point in time, sprawl is manifested in the built-up area:
 - Multiple urban cores
 - Ribbon or strip developments
 - Scatter developments
 - Fragmented and unusable open space
- ▶ Manifestations are quantified with the following metrics:
 - Main core
 - Secondary core(s)
 - Urban fringe
 - Ribbon development
 - Scatter development
- ▶ In sprawling cities, low density developments such as ribbon and scatter are expected to increase over time.

Sprawl Manifestation Metrics Derivation

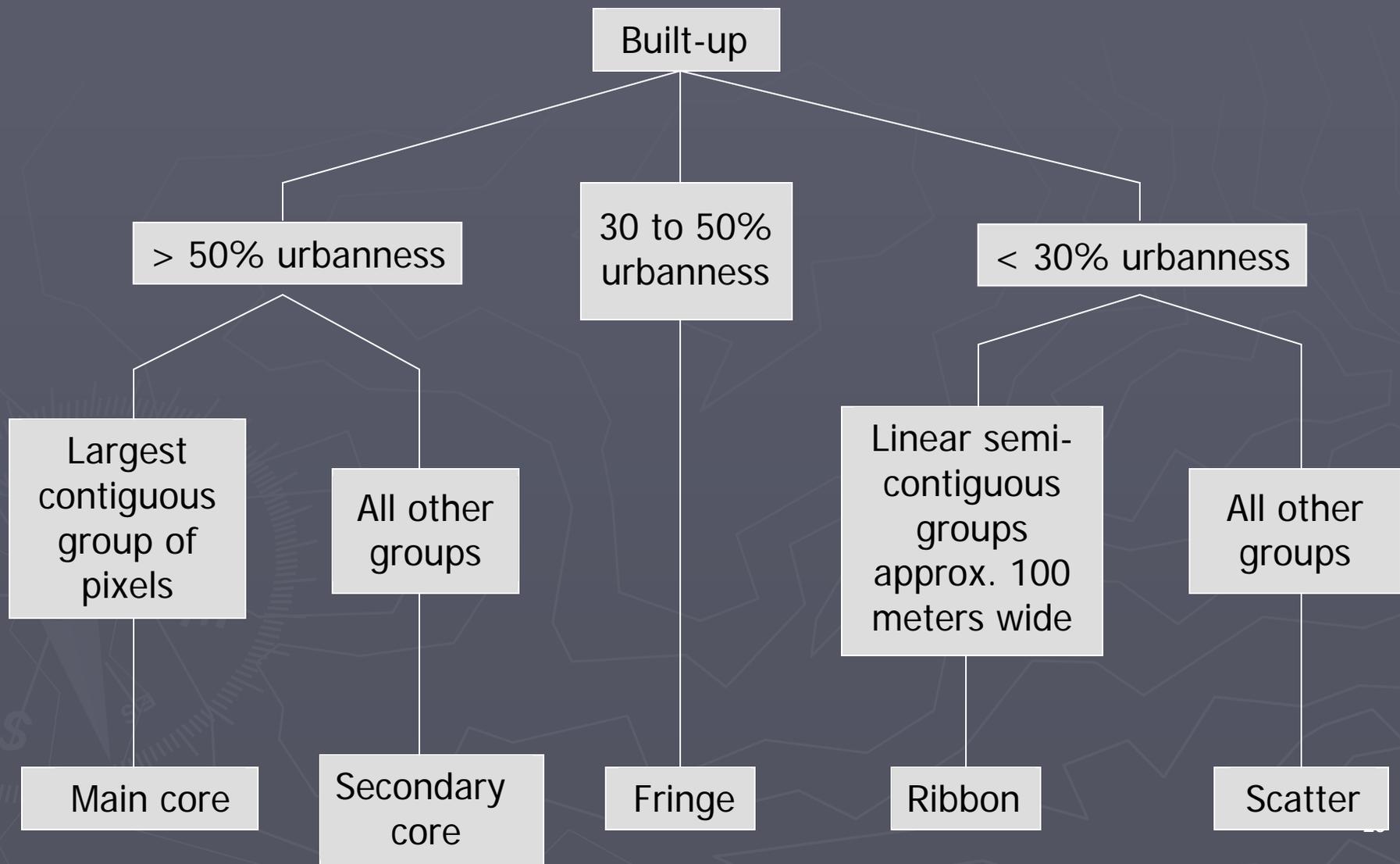
- ▶ Manifestation metrics based on the “urbanness” of a given built-up pixel.
 - Urbanness = percent of neighborhood that is built-up
 - The neighborhood is a 1 km² circle.



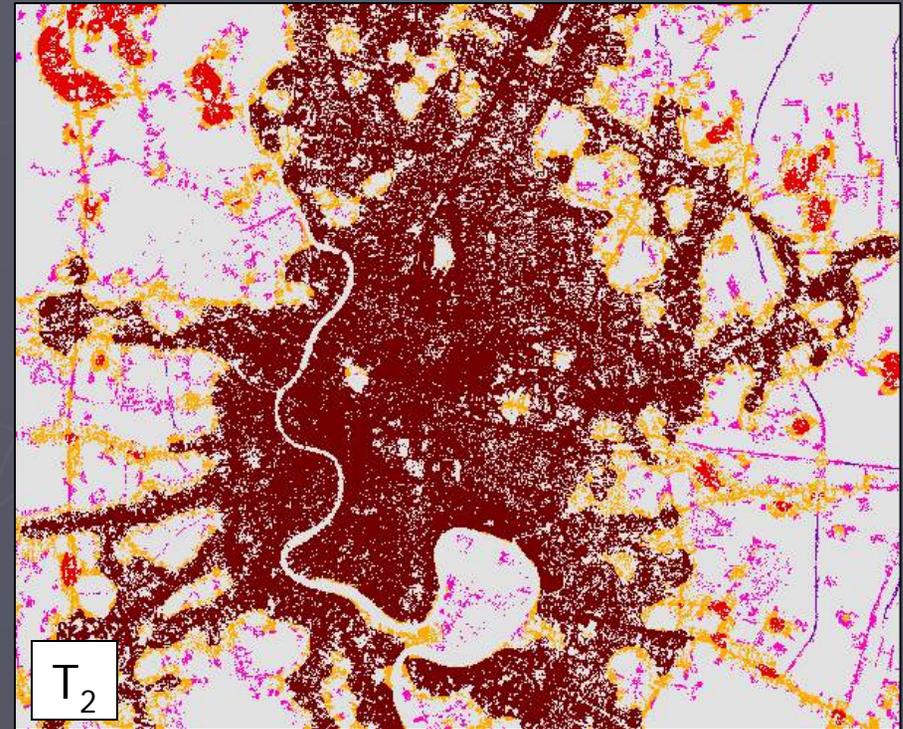
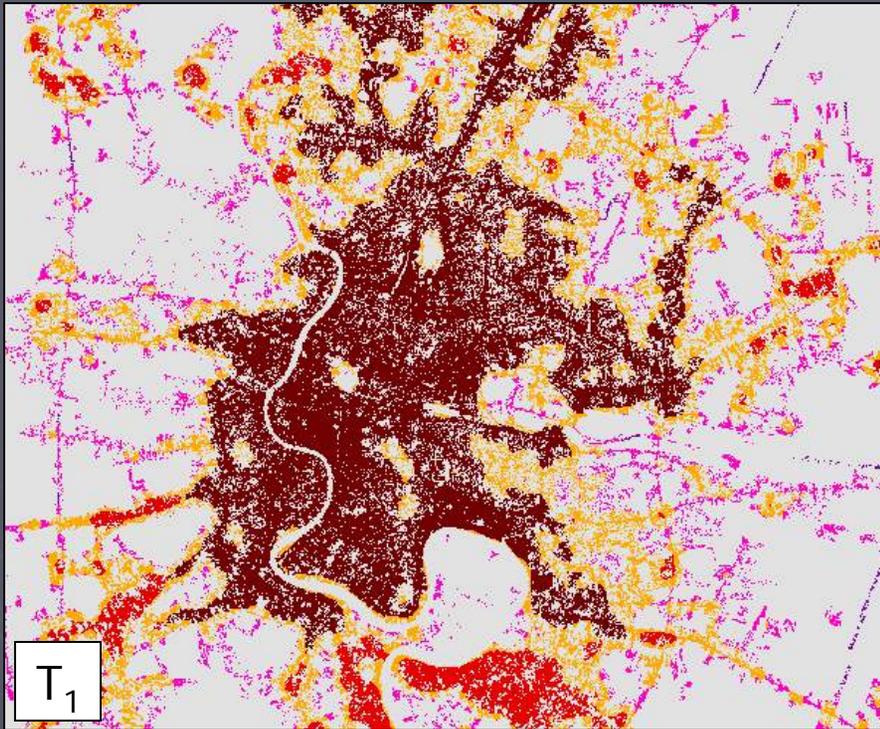
252 built-up pixels = 0.2 km²

urbanness = $0.2 / 1 = 20\%$

Sprawl Manifestation Metrics Definitions



Sprawl Manifestations for Bangkok, Thailand



Sprawl Manifestation Metrics for Bangkok (1994-2002) and Minneapolis (1992-2001)

Metric		T1		T2		Annual ΔT	
		Bangkok	Minneapolis	Bangkok	Minneapolis	Bangkok	Minneapolis
Built-Up Area	km ²	683.0	885.6	1025.9	1100.0	7.0%	2.8%
	%	100%	100%	100%	100%	0.0%	0.0%
Main Core	km ²	211.8	402.5	461.2	555.7	16.3%	4.3%
	%	31.0%	45.5%	45.0%	50.5%	1.9%	0.6%
Secondary Cores	km ²	80.6	94.7	114.2	172.5	5.8%	9.4%
	%	11.8%	10.7%	11.1%	15.7%	-0.1%	0.6%
Urban Fringe	km ²	201.1	234.6	245.6	232.6	-3.1%	-0.0%
	%	29.5%	26.5%	23.9%	21.2%	-0.8%	-0.6
Urban Ribbon	km ²	11.28	8.2	14.2	9.74	3.5%	2.1%
	%	1.7%	0.9%	1.4%	0.9%	-0.0%	0.0%
Urban Scatter	km ²	178.2	145.6	191.1	129.2	1.0%	-0.1%
	%	26.1%	16.4%	18.6%	11.8%	-1.0%	-0.5%

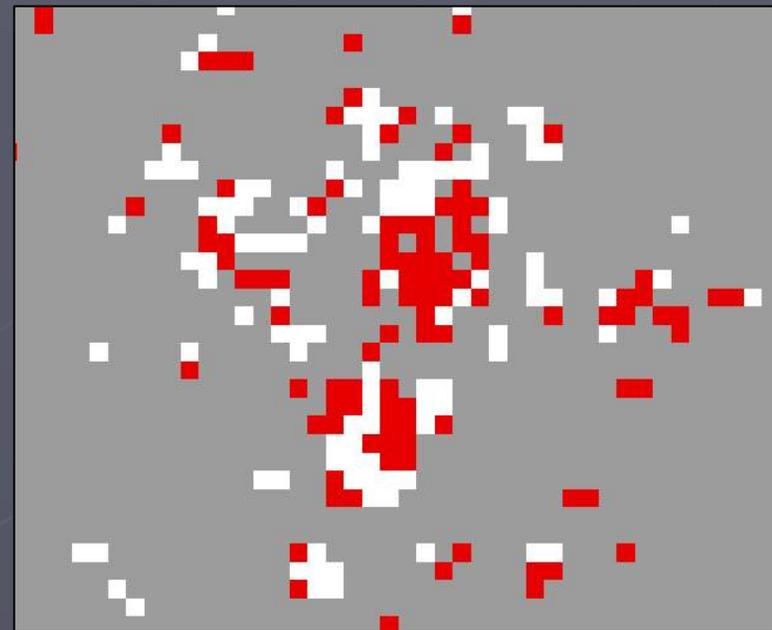
New Development Classification

- ▶ Over a period of time, sprawl is manifested in *new developments* that either extend outward from existing development or 'leapfrog' away from existing development.
- ▶ *New development* is any built-up area added between two time periods - T_1 and T_2 .
- ▶ Classification of new development is based on location relative to the T_1 urban area.

New Development: Infill

Infill:

- ▶ New development occurring within the T_1 urbanized open space
- ▶ Increases contiguity of the built-up area by filling in the urbanized open space



T1 built-up pixels in gray,
infill in red

New Development: Extension

Extension:

- ▶ New non-infill development intersecting the T_1 urban footprint
- ▶ Extends outward from previous development

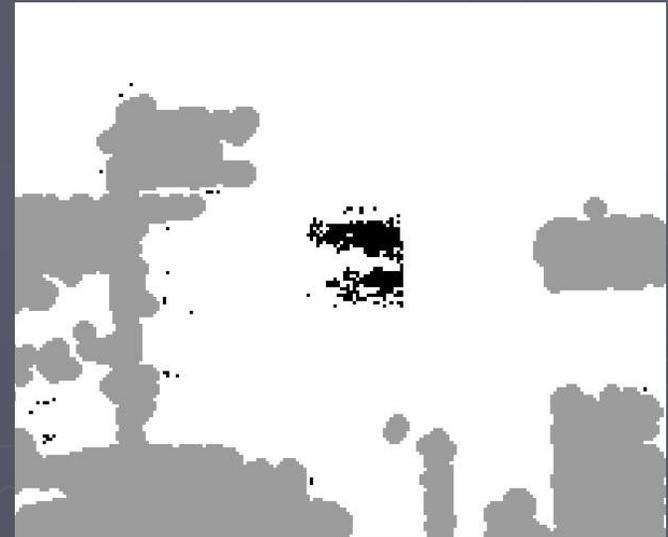


T1 urbanized area in gray,
extension in orange

New Development: Leapfrog

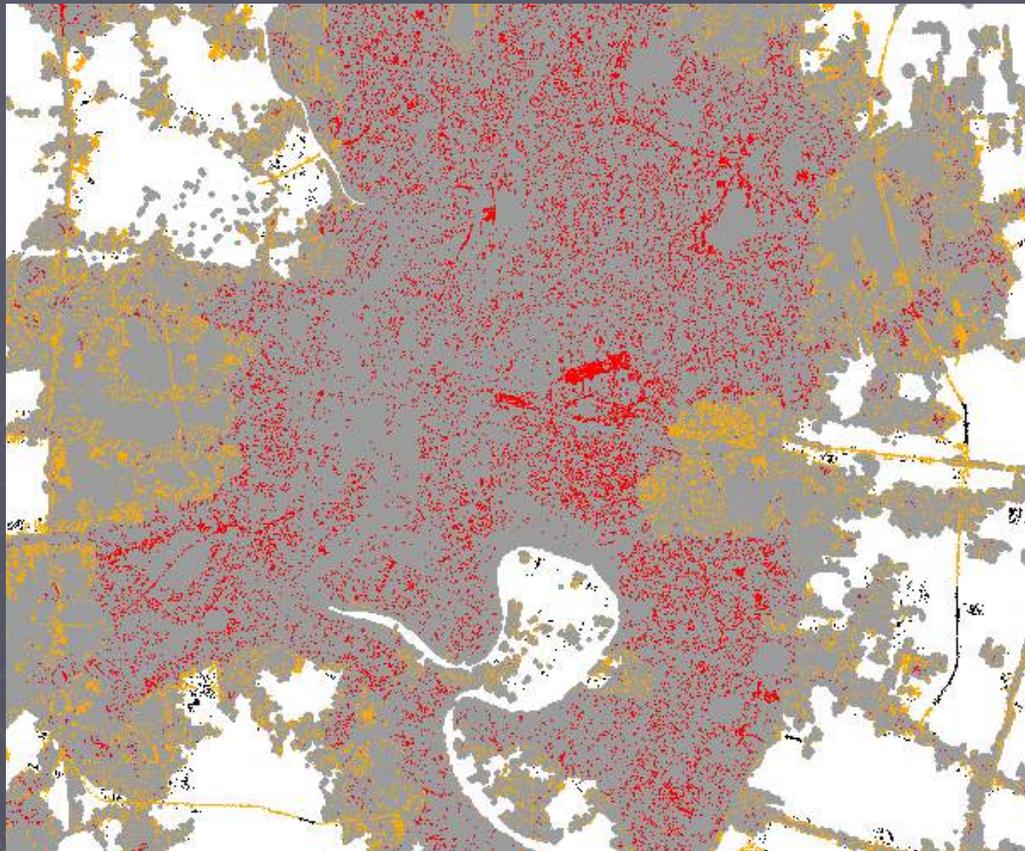
Leapfrog:

- ▶ New development not intersecting the T_1 urban footprint
- ▶ No contiguity with the previous developed areas
- ▶ Has the greatest effect on growth of urban footprint and fragmentation of open lands.



T1 urban footprint in gray,
leapfrog in black

New Development (1994-2002) in Bangkok, Thailand



New Development Metrics for Bangkok (1994-2002) and Minneapolis (1992-2001)

Metric		Bangkok	Minneapolis
New development	km ²	342.89	214.00
Infill	km ²	91.81	80.0
	%	27.0%	37.0%
Extension	km ²	215.1	130.4
	%	63.0%	61.0%
Leapfrog	km ²	36.0	3.8
	%	11.0%	2.0%

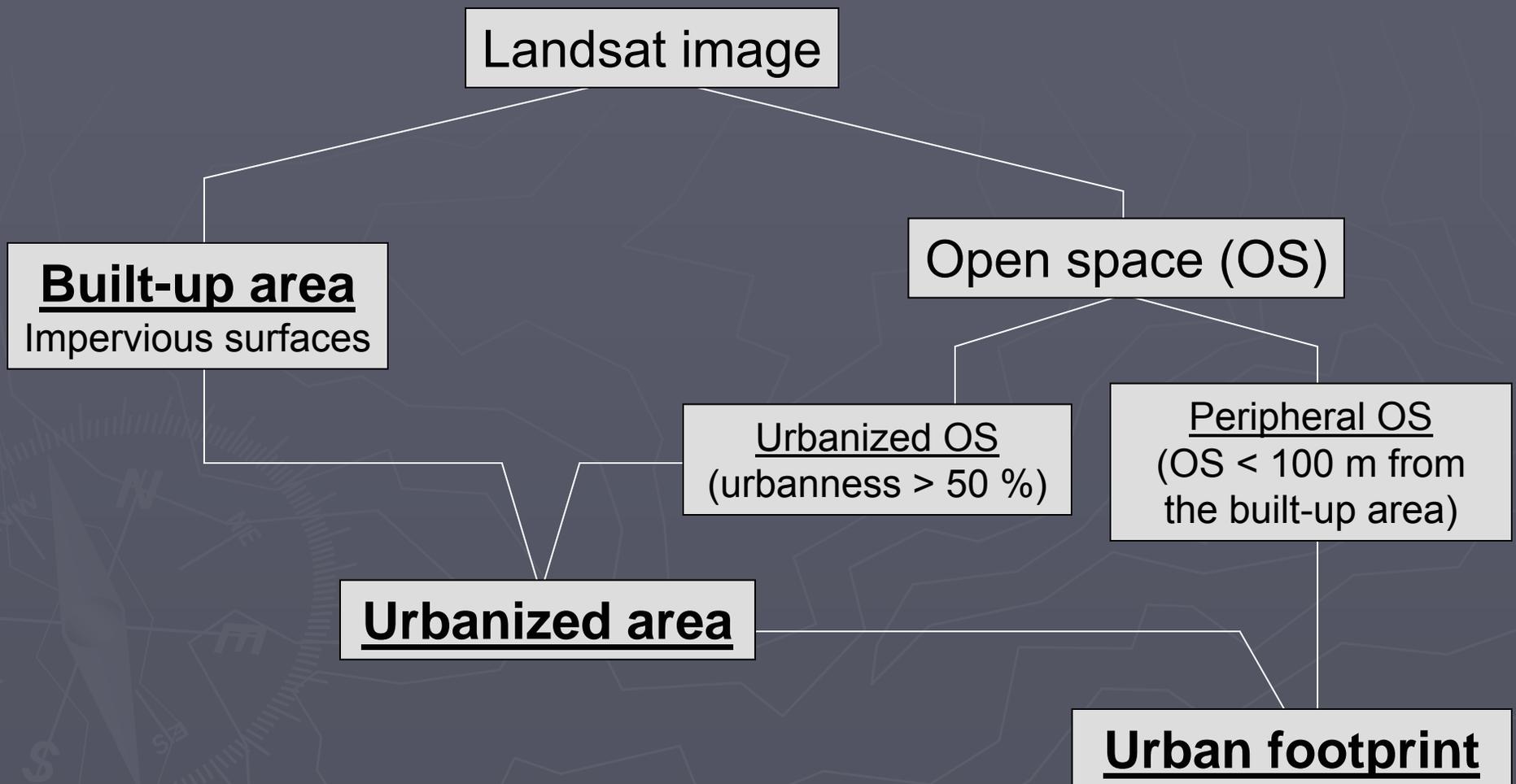
Attributes of Urban Sprawl

- ▶ Sprawl attributes are quantifiable for each city
- ▶ A set of metrics was developed to describe each attribute
 - Metrics in each set tend to be highly correlated
- ▶ We identified four attributes of urban spatial structure that are commonly associated with 'sprawl'...

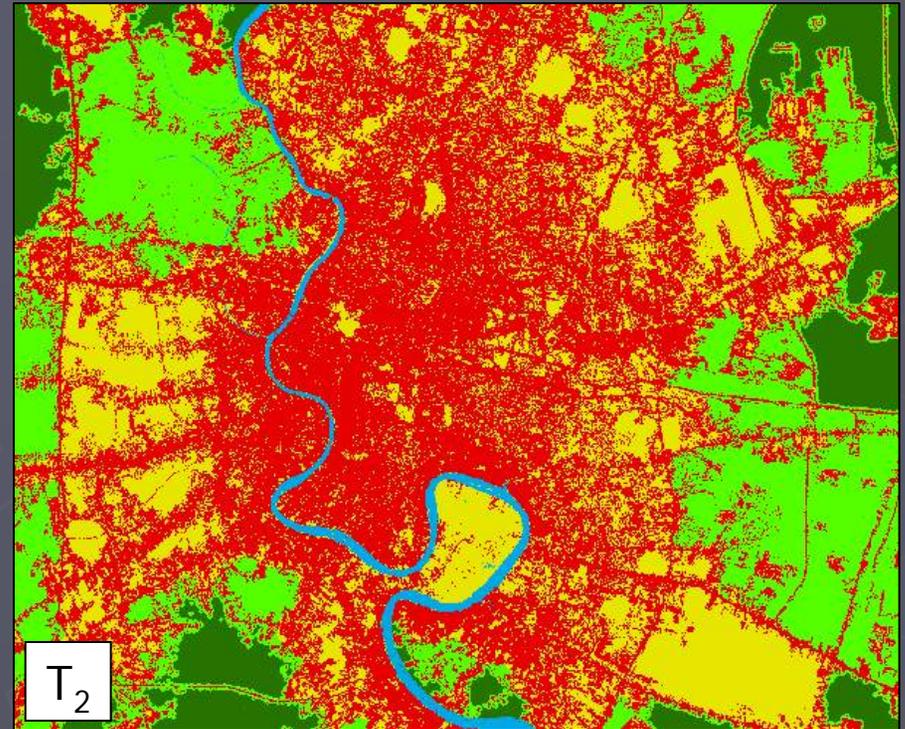
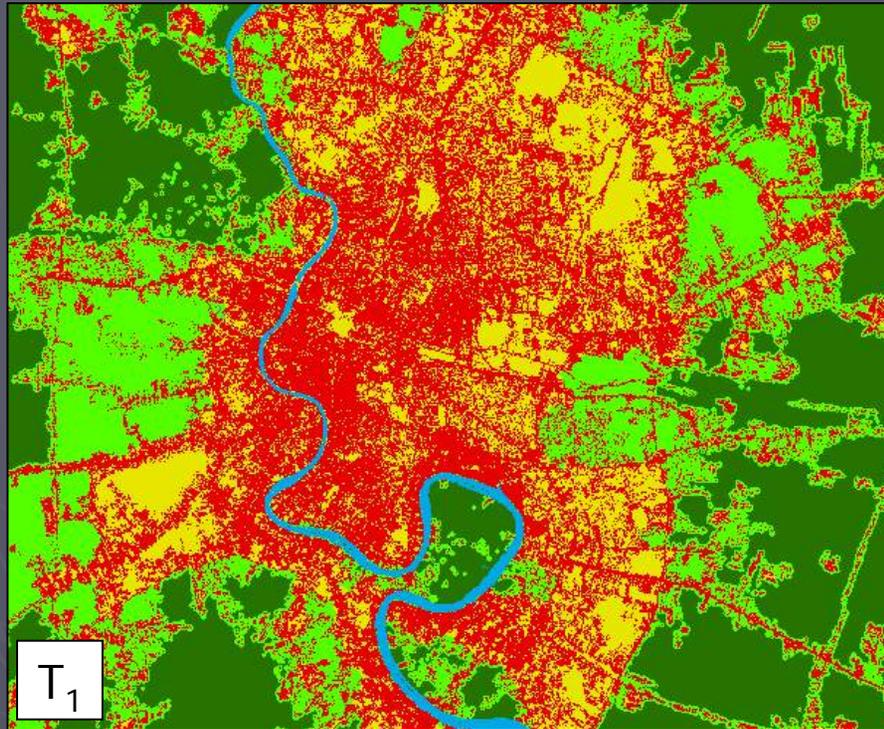
Attribute 1: "Limitless" Urban Extent

- ▶ The land to be considered as urban depends on the issue of interest.
 - Impervious surfaces alone for water quality issues
 - Heavily impacted open land as well as impervious surfaces for ecological issues.
- ▶ We present three measures of urban extent, each corresponding to a different perception of what constitutes a city:
 - Built-up area – impervious surfaces
 - Urbanized area – impervious surfaces as well as heavily influenced open space
 - Urban footprint – impervious surfaces and any open space likely to be impacted, to some degree, by the city.

"Limitless" Urban Extent



Urban Extents for Bangkok, Thailand



Urban Extent Metrics for Bangkok (1994-2002) and Minneapolis (1992-2001)

Metric		T1		T2		Annual ΔT	
		Bangkok	Minneapolis	Bangkok	Minneapolis	Bangkok	Minneapolis
Built-up area	km ²	683.1	886.2	1026.1	1100.0	7.0%	2.7
	%	100%	100%	100%	100%	0.0%	0.0%
Urbanized area	km ²	949.1	1331.5	1456.5	1715.7	5.3%	3.3%
	%	139.0%	150.3%	142.0%	156.0%	0.4%	0.6%
Urbanized OS	km ²	266.0	445.8	430.5	616.0	8.6%	4.4%
	%	39.0%	50.3%	42.0%	56.0%	0.4%	0.6%
Urban Footprint	km ²	1817.9	2314.2	2526.1	2494.7	5.4%	0.9%
	%	266.2%	261.3%	246.2%	226.9%	-2.8%	-3.9%
Peripheral OS	km ²	868.8	982.7	1069.6	779.0	3.2%	-2.4%
	%	127.2%	111.0%	104.3%	70.8%	-3.2%	-4.6%
Affected OS	km ²	1134.9	1428.5	1500.1	1395.0	4.4%	-0.3%
	%	166.2%	161.2%	146.2%	126.8%	-2.8%	-3.9%

Attribute 2: Declining Population Density

- ▶ Population densities are expected to decline over time in sprawling cities.
- ▶ Population densities measured for the three types of urban extent.

Population Density Metrics for Bangkok (1994-2002) and Minneapolis (1992-2001)

Metric (Densities in persons / ha)	T1		T2		Annual ΔT	
	Bangkok	Minneapolis	Bangkok	Minneapolis	Bangkok	Minneapolis
Population	8,245,332	2,166,839	9,768,215	2,483,342	2.4%	1.6%
Built-up area density	121	24	95	23	-3.2%	-0.9%
Urbanized area density	87	16	67	14	-3.5%	-1.3%
Urban footprint density	45	9	39	10	-2.2%	0.7%

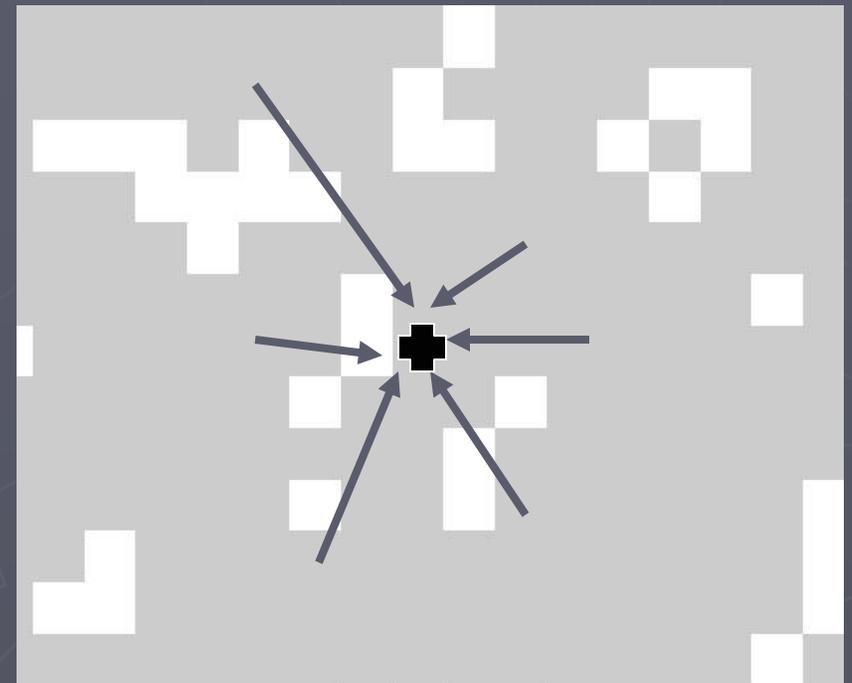
Attribute 3: Non-Compactness of the Urban Area

- ▶ The compactness of a sprawling city is expected to decline over time due to leapfrogging and low density development.
- ▶ We measure compactness of the *urbanized area*.
- ▶ Metrics based on assumption that the circle is the most compact shape possible for a given area.
- ▶ Compactness metrics are normalized and range from 0 to 1 with higher values indicating more compact shapes.

Measuring Compactness: The Proximity Index

The Proximity Index:

- ▶ Based on average distance, of all points in the urbanized area, to the center of gravity (centroid) of the urbanized area.
- ▶ Normalized by the average distance (d) to centroid of the *equal area circle* with radius (r) – a circle with an area equal to that of the urbanized area.
 - $d = (2 / 3) * r$

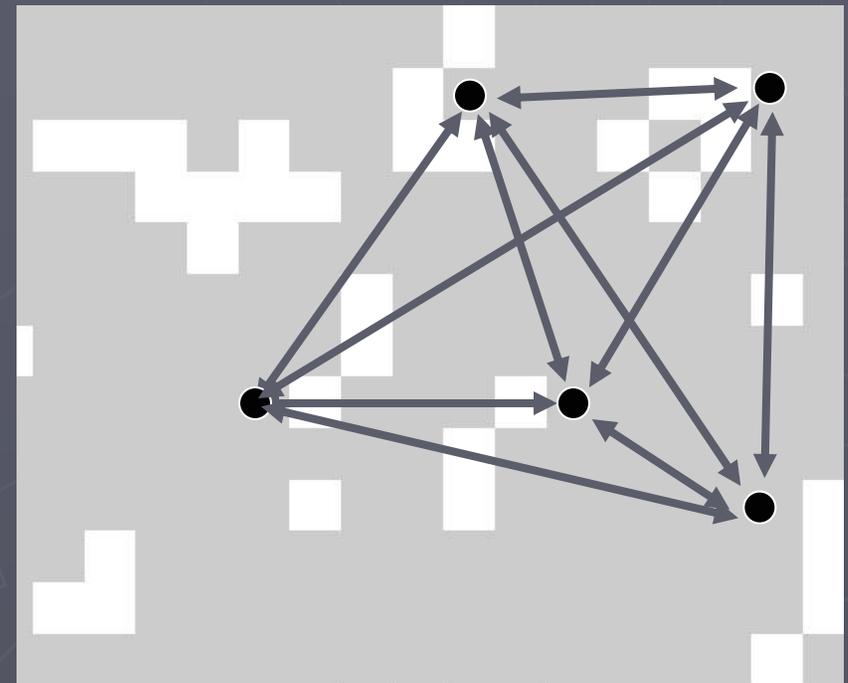


Urbanized area in gray,
centroid in black

Measuring Compactness: The Cohesion Index

The Cohesion Index:

- ▶ Based on the average distance between all possible pairs of points in the urbanized area
- ▶ Normalized by average distance (d) between all pairs of points within the *equal area circle* with radius (r).
 - $d = 0.9054 * r$



Urbanized area pixels in gray,
sample points in black

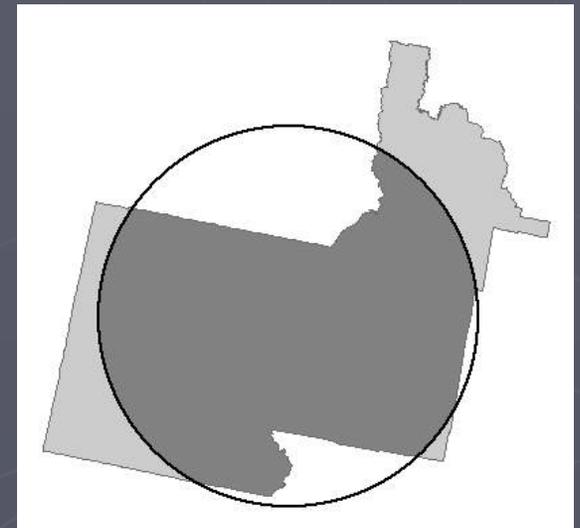
Measuring Compactness: The Compactness and Constrained Compactness Indices

The Compactness Index

- ▶ The fraction of the shape area that is within an *equal area circle* centered at the shape centroid.

The Constrained Compactness Index

- ▶ The fraction of the shape area that is within the *net equal area circle* centered at the shape centroid.
- ▶ The *net equal area circle* has a buildable area (excluding water and excessive slope) equal to the urbanized area



Compactness = 0.75

Compactness Metrics for Bangkok (1994-2002) and Minneapolis (1992-2001)

Metric	T1		T2		Annual ΔT	
	Bangkok	Minneapolis	Bangkok	Minneapolis	Bangkok	Minneapolis
Proximity	0.67	0.87	0.75	0.91	1.11%	0.46%
Cohesion	0.65	0.85	0.72	0.89	0.97%	0.46%
Compactness	0.62	0.75	0.66	0.80	0.56%	0.57%
Constrained compactness	0.62	0.77	0.67	0.82	0.62%	0.57%

Attribute 4: Non-contiguity of the Built-up Area and the Fragmentation of Open Land

- ▶ Sprawling cities are typically expected to become less contiguous over time while adjacent open land becomes increasingly fragmented.
- ▶ Contiguity metrics are based on the built-up area.
- ▶ Higher values for all metrics indicate lower contiguity of the built-up area and greater fragmentation of open space.

Openness

The Openness Index:

- ▶ The average *openness* of all built-up pixels.
- ▶ *Openness* is the fraction of open land within a 1 km² neighborhood around a given built-up pixel – the complement of *urbanness*.
- ▶ Values range from 0 to 1



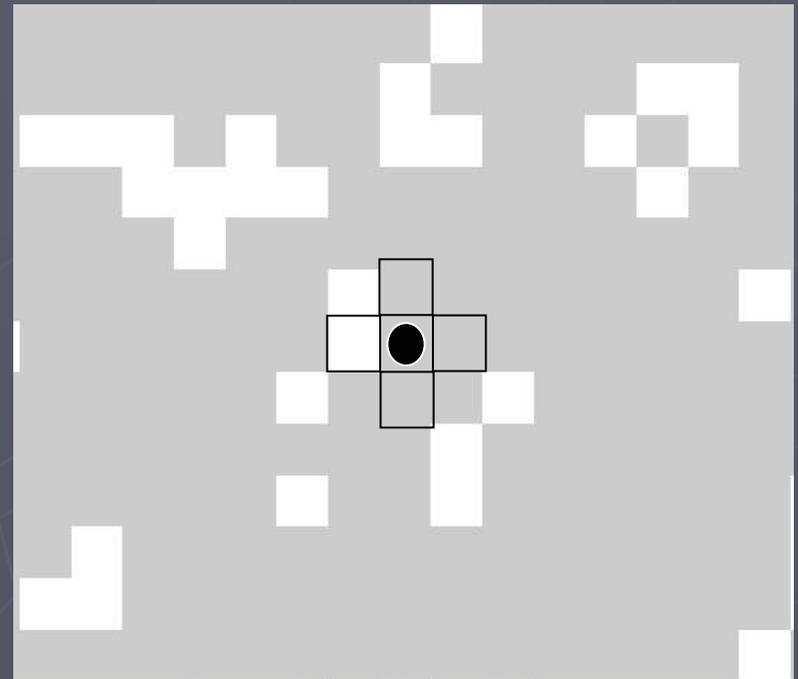
252 built-up pixels = 0.2 km²

$$\text{openness} = \frac{1 \text{ km}^2 - 0.2 \text{ km}^2}{1 \text{ km}^2} = 0.8$$

Open Space Contiguity

The Open Space Contiguity Index:

- ▶ The fraction of built-up pixels that are cardinally adjacent to at least one open space pixel.
- ▶ Calculated using a neighborhood operation
- ▶ Values range from 0 to 1



Built-up pixels in gray
Open space pixels in white

Open Space Fragmentation

The Open Space Fragmentation Index:

- ▶ The ratio of the *affected open land* to the built-up area.
- ▶ *Affected open lands* include the urbanized open space and the peripheral open space
- ▶ Values are greater than zero

Contiguity and Openness metrics for Bangkok (1994-2002) and Minneapolis (1992-2001)

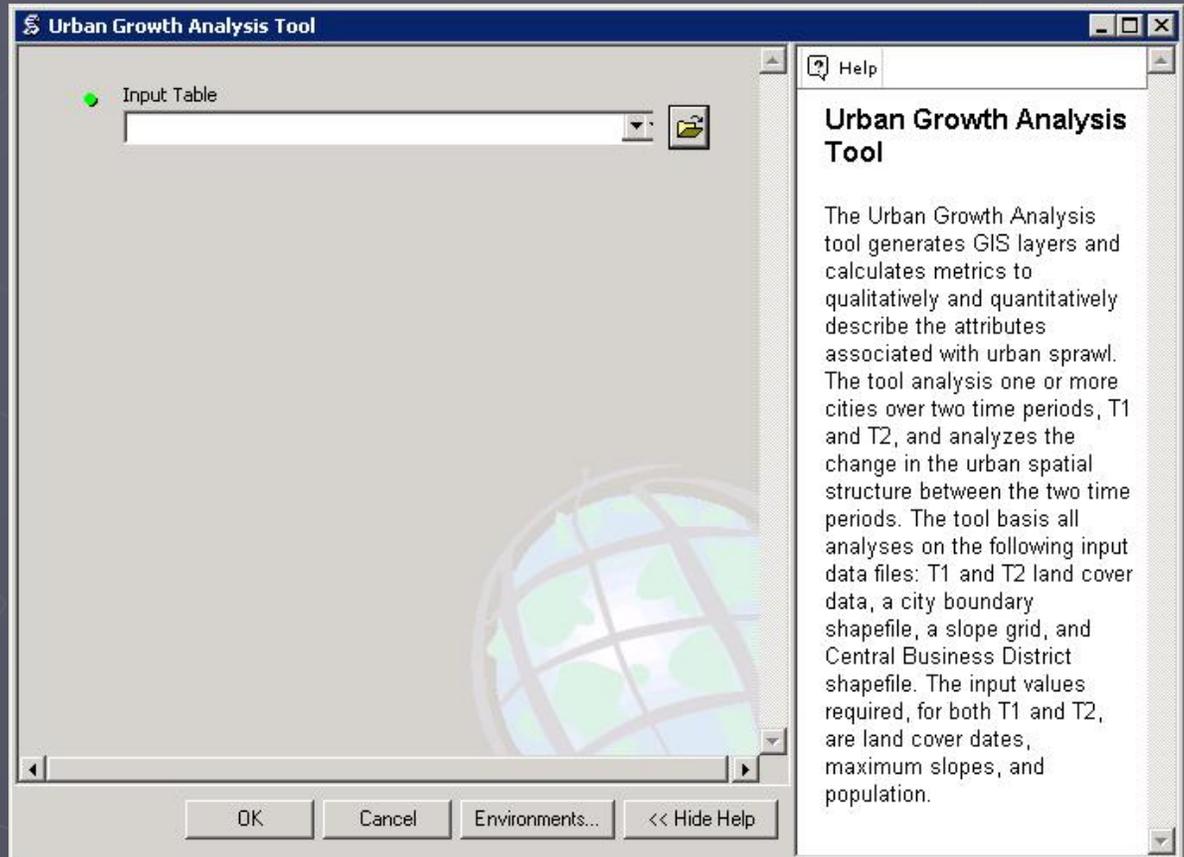
Metric	T1		T2		Annual ΔT	
	Bangkok	Minneapolis	Bangkok	Minneapolis	Bangkok	Minneapolis
Openness Index	0.54	0.46	0.46	0.41	-1.10%	-0.55%
Open Space Contiguity	0.56	0.55	0.47	0.50	-1.00%	-0.60%
Open Space Fragmentation Index	1.66	1.61	1.46	1.27	-2.80%	-3.90%

The Urban Growth Analysis Tool (UGAT)

- ▶ Python script that works with ArcGIS 9.2
 - Executable through python or ArcGIS toolbox
- ▶ UGAT performs all analyses needed to derive the metrics and create GIS layers
- ▶ Data is input into the tool via a table
 - Table may contain data for any number of cities
 - UGAT will run the analysis for each city listed in the table

The Urban Growth Analysis Tool

ArcToolbox Interface



The Urban Growth Analysis Tool Python Interface

```
1 # Name: Jason Parent
2 # Date: March 2008
3 # Purpose: 1) Create raster layers depicting manifestations of urban spread for two time
4 #           2) Calculate metrics to quantify compactness and change in compactness of urb
5
6
7 # Import system modules...
8 import Helper_UGA,time,sys,os,math
9
10 # Create the geoprocessor object...
11 gp = Helper_UGA.gpcreate()
12
13 ##inputDataTable = sys.argv[1]
14 inputDataTable = r"C:\Worldbank\Pathnames.mdb\InputDataTable"
15
16 ##DataWS = sys.argv[2]
17 DataWS = r"C:\Worldbank\Input_Data"
18 #DataWS = r"C:\Projects\CT_VGA\Input_data"
19
20 ##OutputWS = sys.argv[3]
21 #OutputWS = r"C:\Worldbank\Output_Data"
22 OutputWS = r"C:\UGA_test"
23
24 errorFile = "%s\UGA_errFile.txt" % OutputWS
25
26 # name of project temp folder...
27 TempFolder = "C:\UGA_Temp"
28
29 # temporary workspace for ESRI files...
30 TempWS = "%s\UGA_temp.gdb" % TempFolder
31
32
33 # get workspaces for all input data files...
34 wkspDict = Helper_UGA.GetWksp (DataWS)
35
36 #-----
37 # URBAN LANDSCAPE CATEGORY VALUES...
38
39 core, fring, ribbon, scatter = 1, 2, 3, 4
40 urbOS, i_urbOS, watOS, slpOS = 5, 6, 7, 8
41 periOS, i_periOS, watPOS, slpPOS = 9, 10, 11, 12
```

Conclusions

- ▶ Phase II is currently ongoing as is development of the Urban Growth Analysis Tool. Future papers will provide in-depth and comprehensive discussions of the metrics developed in this project
- ▶ The metrics, developed so far in this project, allow rigorous quantitative assessment of the change in urban spatial structure over time
- ▶ Metrics in a set tend to be highly correlated
 - provides alternative ways to measure each attribute.
- ▶ The Urban Growth Analysis Tool makes it practical to perform the analysis for a large number of cities. The tool will be made available to researchers interested in applying the analysis to their own study areas.

References

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QUESTIONS?

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