

# Urban Forestry and Remote Sensing

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# Cities

- United Nations, Istanbul , 2001, p. 1
  - Cities draw people
    - Services, resources, etc.
  - 1800 – only 2% lived in cities/urban centers
  - Currently ~ 6.4 billion – slightly less than  $\frac{1}{2}$  live in cities/urban centers
  - 2007 – over  $\frac{1}{2}$  will live in cities/urban centers
  - Demographic transformation

# Urban Areas

- United States
  - 80% of the population
  - 25% of the nation's total tree canopy cover



# Why Study?

- 1970-1990 urban density *decreased* by 23% (AP)
- Changes in land use alter ecosystem
  - Structure
  - Function
  - Pattern
- Alter local climate



<http://www.ci.atlanta.ga.us/skyline.htm>

# Major US Cities

| 1999 rank | 1990 rank | Name                    | Population<br>1999 (est.) | Population<br>1990 | % change |
|-----------|-----------|-------------------------|---------------------------|--------------------|----------|
| 1         | 1         | New York,<br>N.Y.       | 7,428,162                 | 7,322,564          | 1.40%    |
| 2         | 2         | Los<br>Angeles,         | 3,633,591                 | 3,485,499          | 4.20%    |
| 3         | 3         | Chicago, Ill.           | 2,799,050                 | 2,783,660          | 0.60%    |
| 4         | 4         | Houston,<br>Texas       | 1,845,967                 | 1,697,873          | 8.70%    |
| 5         | 5         | Philadelphia<br>, Pa.   | 1,417,601                 | 1,585,577          | -10.60%  |
| 6         | 6         | San Diego,<br>Calif.    | 1,238,974                 | 1,111,031          | 11.50%   |
| 7         | 10        | Phoenix,<br>Ariz.       | 1,211,466                 | 988,983            | 22.50%   |
| 8         | 9         | San<br>Antonio,         | 1,147,213                 | 997,434            | 15.00%   |
| 9         | 8         | Dallas,<br>Texas        | 1,076,214                 | 1,006,646          | 6.90%    |
| 10        | 7         | Detroit,<br>Mich.       | 965,084                   | 1,027,946          | -6.10%   |
| 11        | 11        | San Jose,<br>Calif.     | 867,675                   | 783,324            | 10.80%   |
| 12        | 14        | San<br>Francisco,       | 746,777                   | 723,959            | 3.20%    |
| 13        | 13        | Indianapolis<br>, Ind.  | 738,907                   | 731,726            | 1.00%    |
| 14        | 16        | Jacksonville<br>, Fla.  | 695,877                   | 635,042            | 9.60%    |
| 15        | 15        | Columbus,<br>Ohio       | 671,247                   | 636,323            | 5.50%    |
| 16        | 12        | Baltimore,<br>Md.       | 632,681                   | 736,014            | -14.00%  |
| 17        | 22        | El Paso,<br>Texas       | 612,770                   | 515,652            | 18.80%   |
| 18        | 18        | Memphis,<br>Tenn.       | 606,109                   | 618,894            | -2.10%   |
| 19        | 25        | Austin,<br>Texas        | 587,873                   | 494,290            | 18.90%   |
| 20        | 17        | Milwaukee,<br>Wisc.     | 572,424                   | 628,300            | -8.90%   |
| 21        | 20        | Boston,<br>Mass.        | 555,249                   | 574,289            | -3.30%   |
| 22        | 21        | Seattle,<br>Wash.       | 537,150                   | 516,332            | 4.00%    |
| 23        | 33        | Charlotte,<br>N.C.      | 520,829                   | 426,984            | 22.00%   |
| 24        | 19        | Washington<br>, D.C.    | 519,000                   | 606,900            | -14.50%  |
| 25        | 26        | Nashville-<br>Davidson, | 506,385                   | 488,188            | 3.70%    |

# Urban Sprawl

- Land use / land cover conversion
- Human habitat
- Urban forest
  - Concern



# Urban Forests

- Trees and other vegetation that grow in urban and suburban areas
  - Public lands
  - Private lands
- Require investments



Source: Alachua County Visitors and Conventions Bureau

# Urban Forests

- Valued for many reasons
  - Pleasant landscape
  - Peace and quiet
  - Screening and privacy
  - Recreation opportunities
  - Increase property values

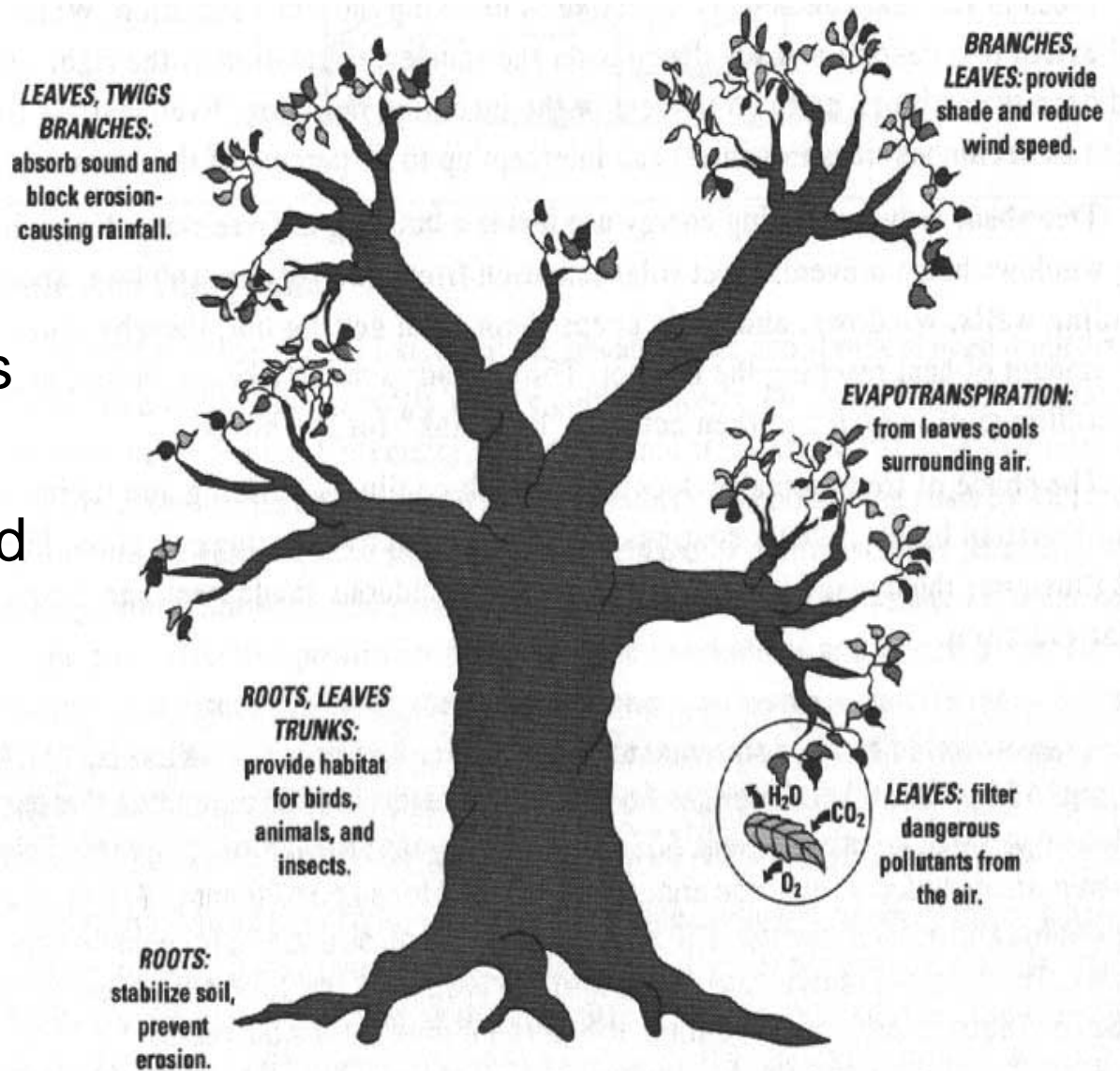


Moeller Realty, Inc.



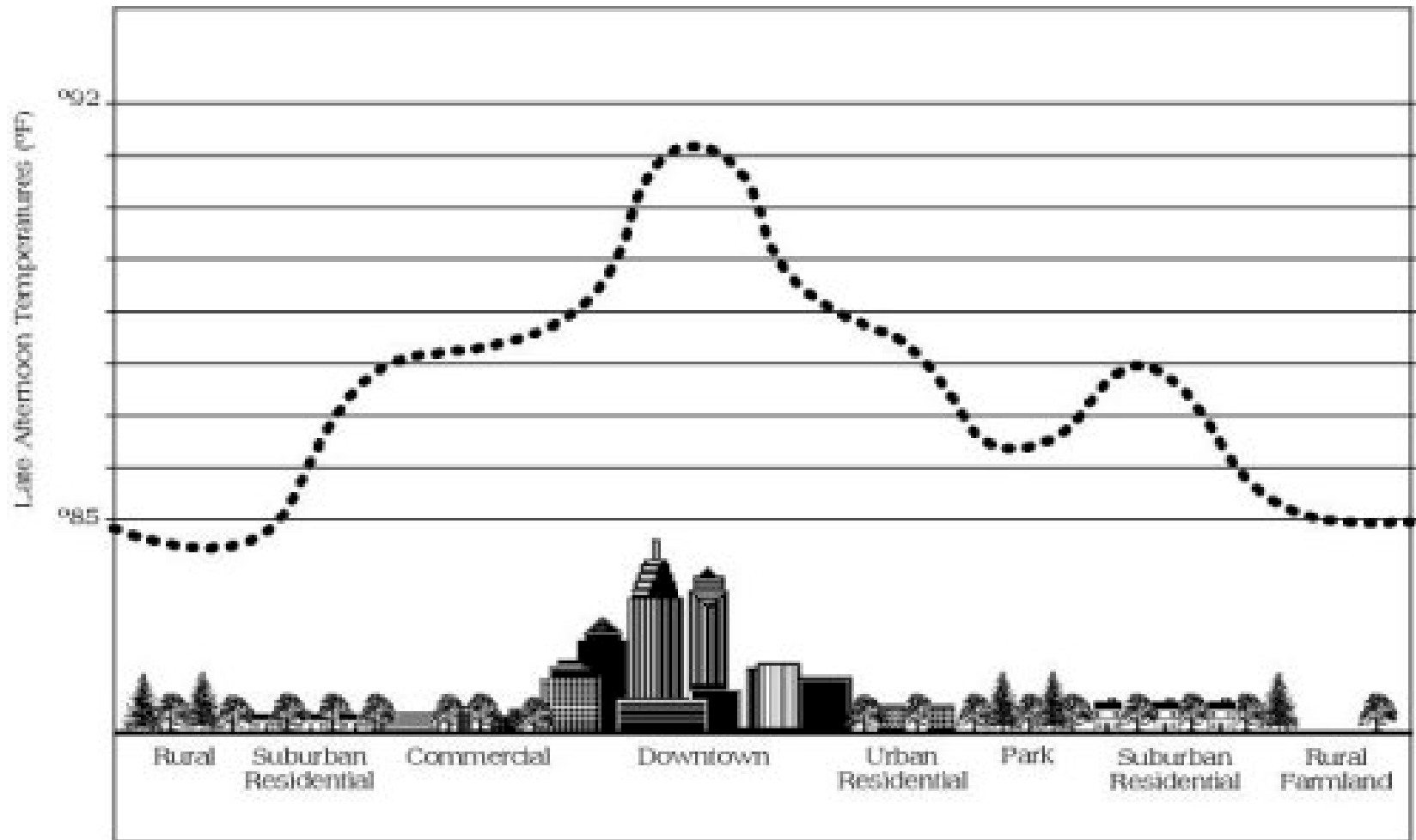
# Environmental Benefits

- High urban leaf areas
  - Air-filtering systems
  - Reduce runoff
  - Reduce energy load
  - Wildlife habitat and diversity
  - Cool air
    - Shade
    - Evapotranspiration



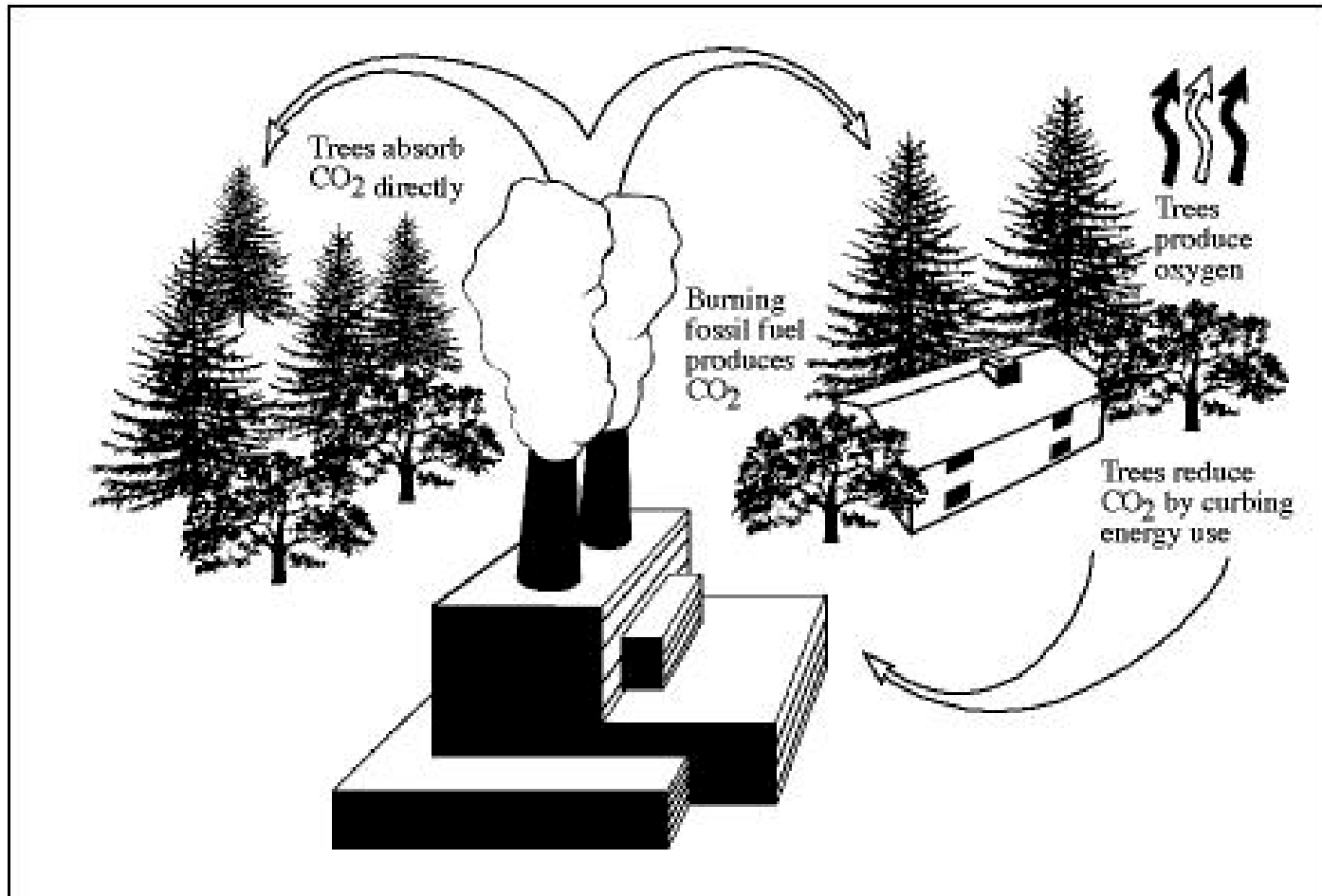
Akbari et al., 1992, p. 27, fig 2-1.

# Urban-Heat Island

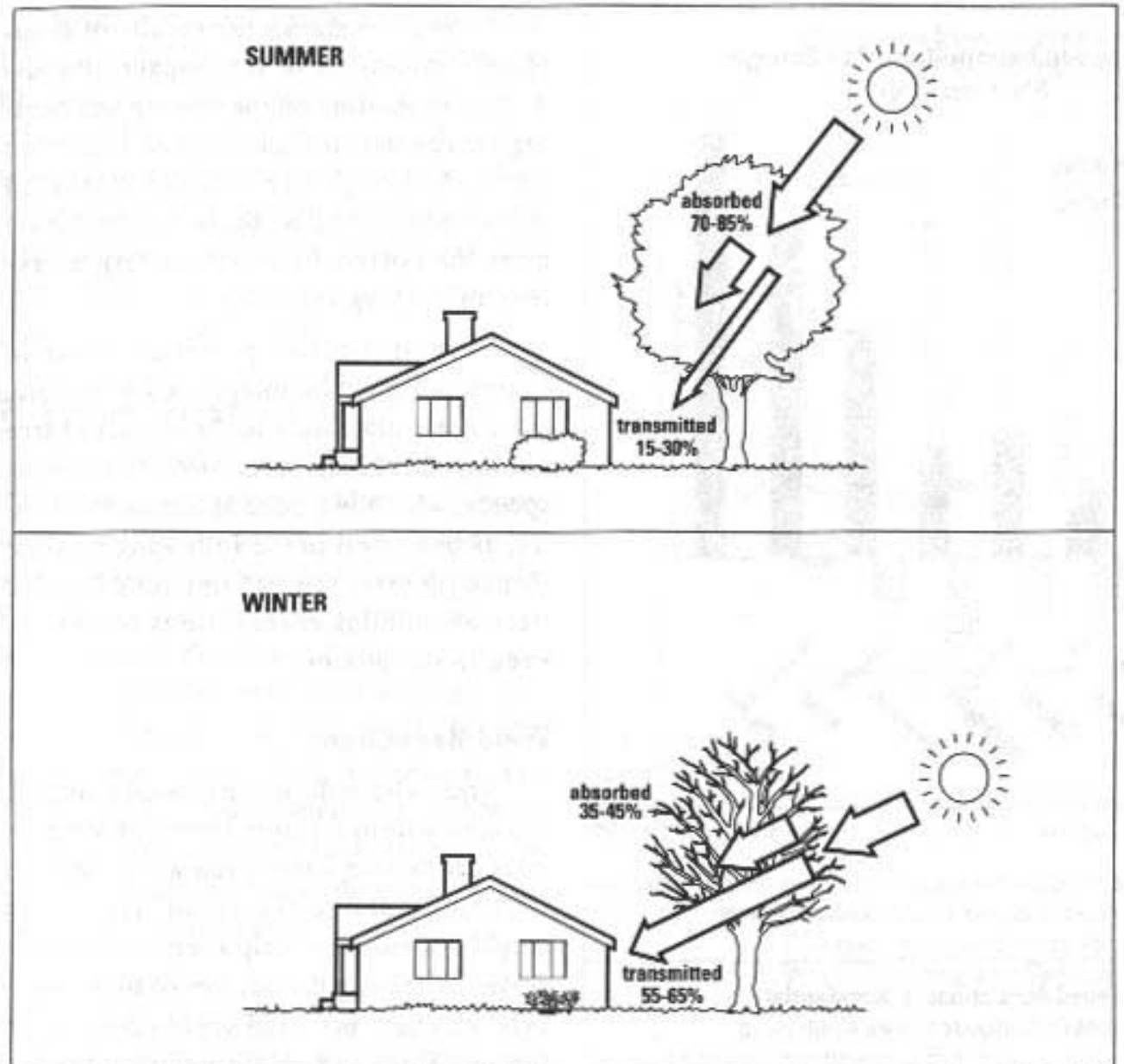


Akbari et al., 1992, p. 9, fig 1-4.

# CO<sub>2</sub> Reduction



# Energy Reduction



Source: Heister, 1996

Akbari et al., 1992, p. 29, fig 2-2.

# Municipal Tree Ordinances

- Necessary
  - Education alone does not guarantee action
- Some residents don't appreciate the value of tree ordinances
- Very difficult to prove \$ saved or gained
  - How to quantify \$ saved

# Disadvantages of Urban Forests

- Fill landfills
  - 20% of municipal waste (EPA)
    - Compost soil amendment
    - Wood chips for fuel
    - Increased shade could decrease ground cover
- Higher water demand – not true
  - Native trees
  - Lawns use more water than trees

# Disadvantage



# Disadvantage





# Disadvantage



# Challenge

- Convince people
- Money
- Penalties
- Geospatial technologies
  - Remote sensing
  - GIS

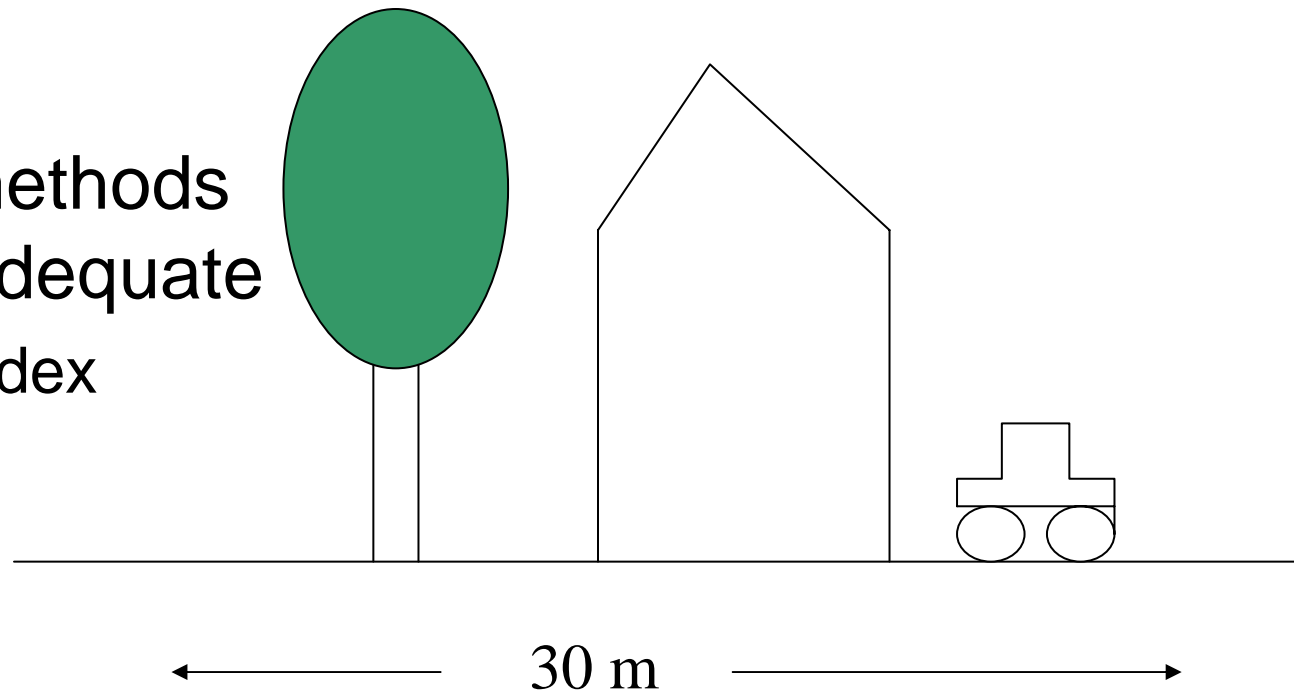


# Urban Remote Sensing

- Implemented by many agencies
  - Planners
  - Tax assessors
  - Transportation
  - Utility companies
  - Departments of parks, recreation, and tourism
- Accuracy

# Urban Remote Sensing

- Careful in interpretation
- Parametric methods may not be adequate
  - Leaf Area Index

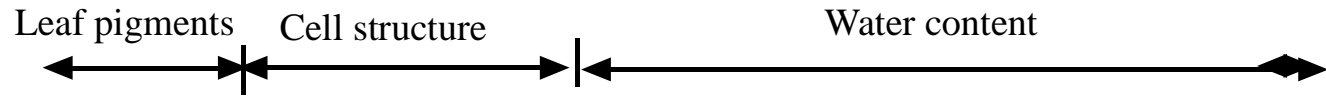


# Leaf Area Index

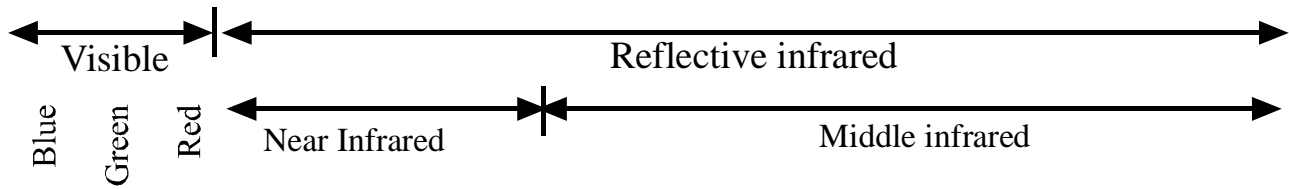
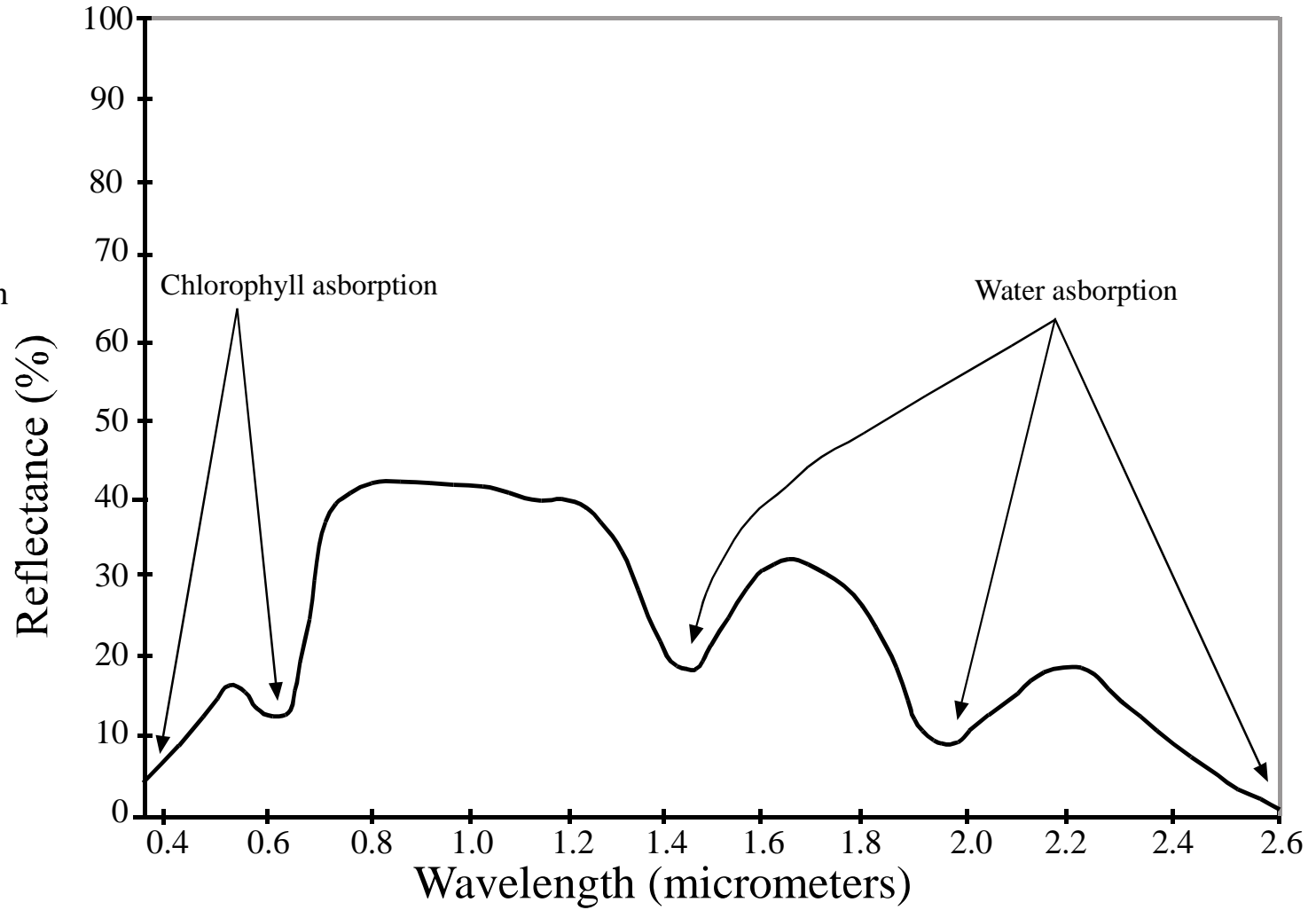
- Useful in ecosystem analysis
- Related to biological processes
  - Primary productivity
  - Canopy gas exchange
- Used in global circulation models
- Estimated via RS



Dominate factor  
controlling leaf  
reflectance

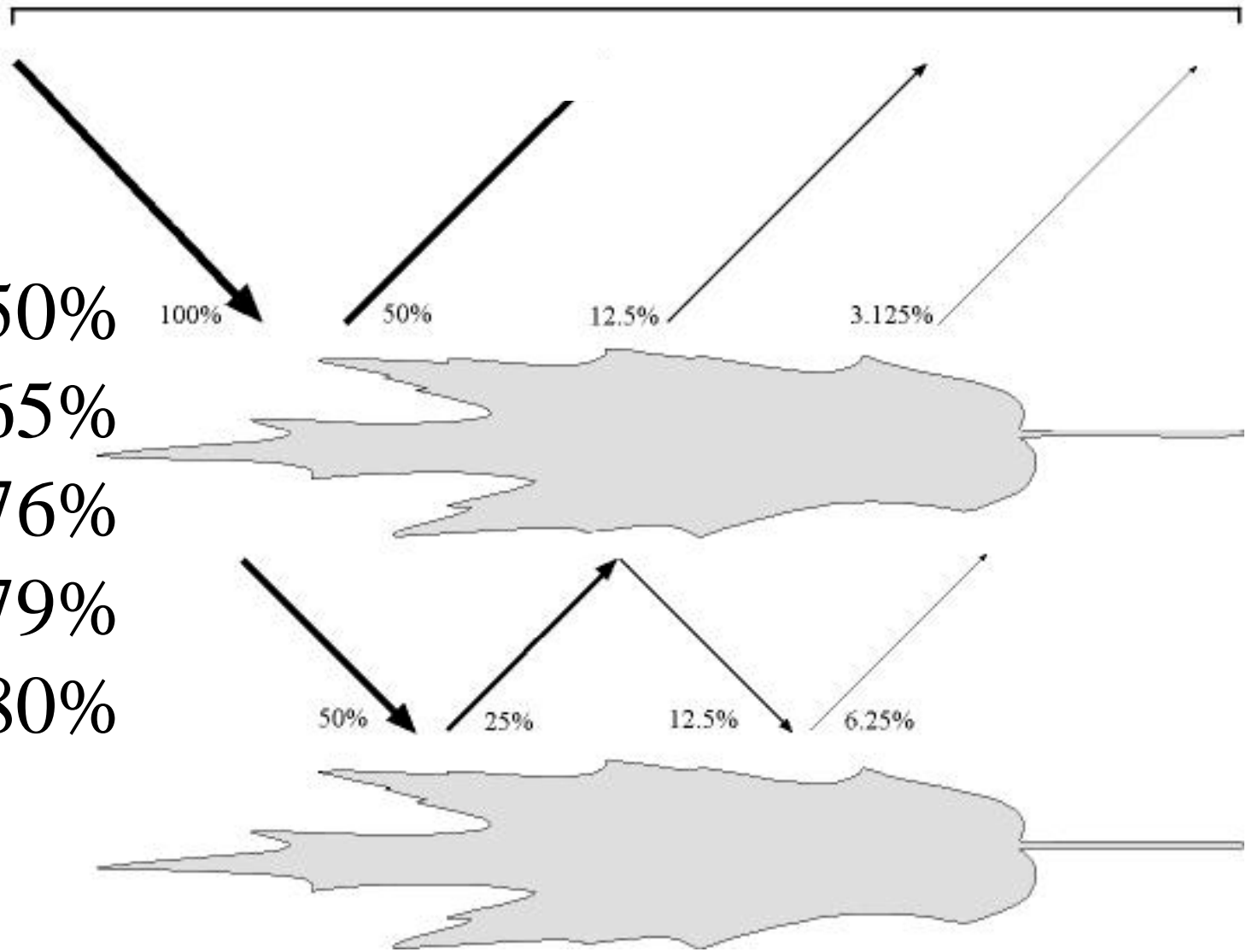


Primary  
absorption  
bands

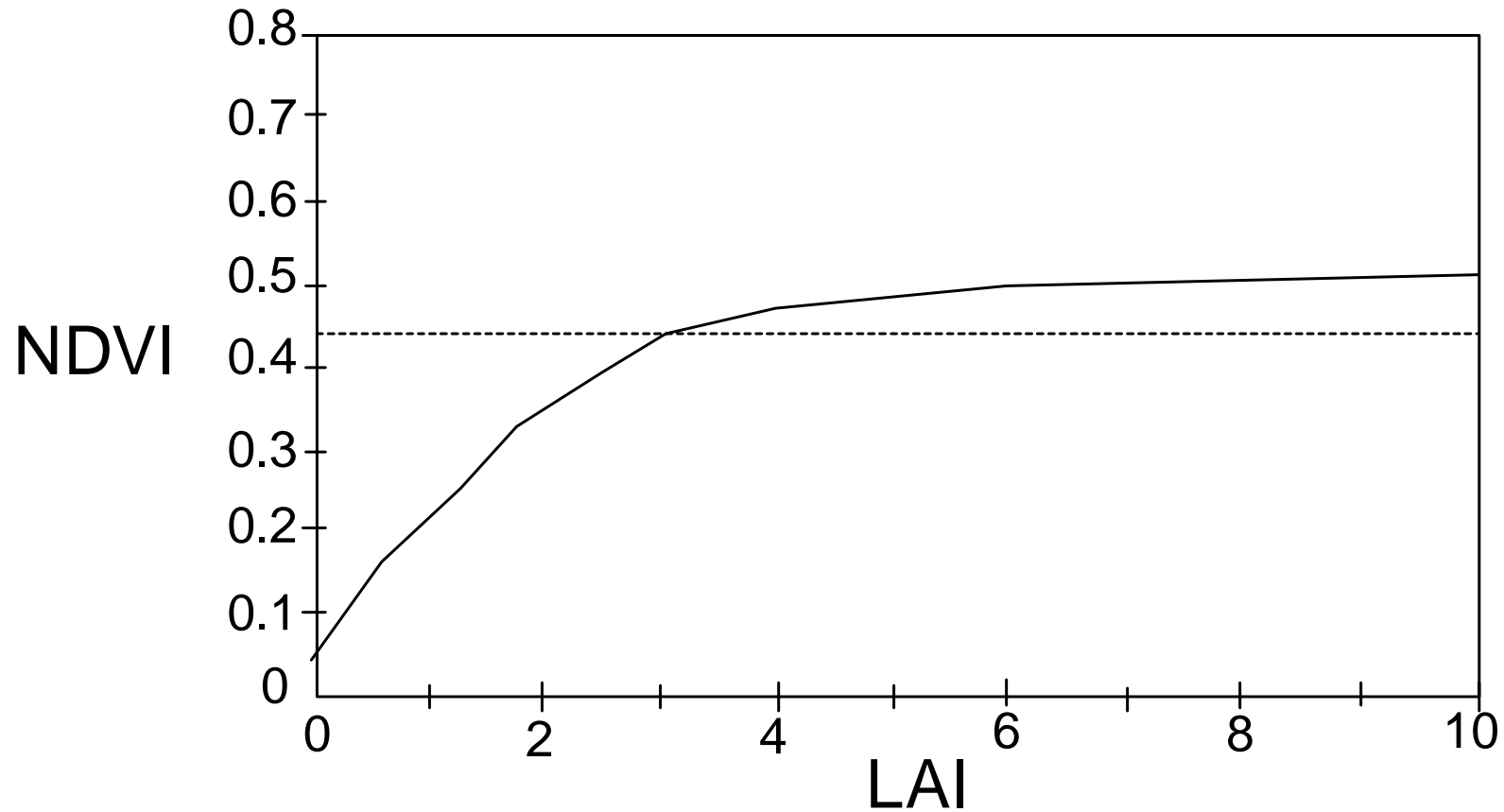


Total NIR reflectance = 65.625% of incident energy

- 1 leaf = 50%
- 2 leaves = 65%
- 3 leaves = 76%
- 4 leaves = 79%
- 5 leaves = 80%



# NDVI Saturation



(adapted from Carlson and Ripley, 1997).



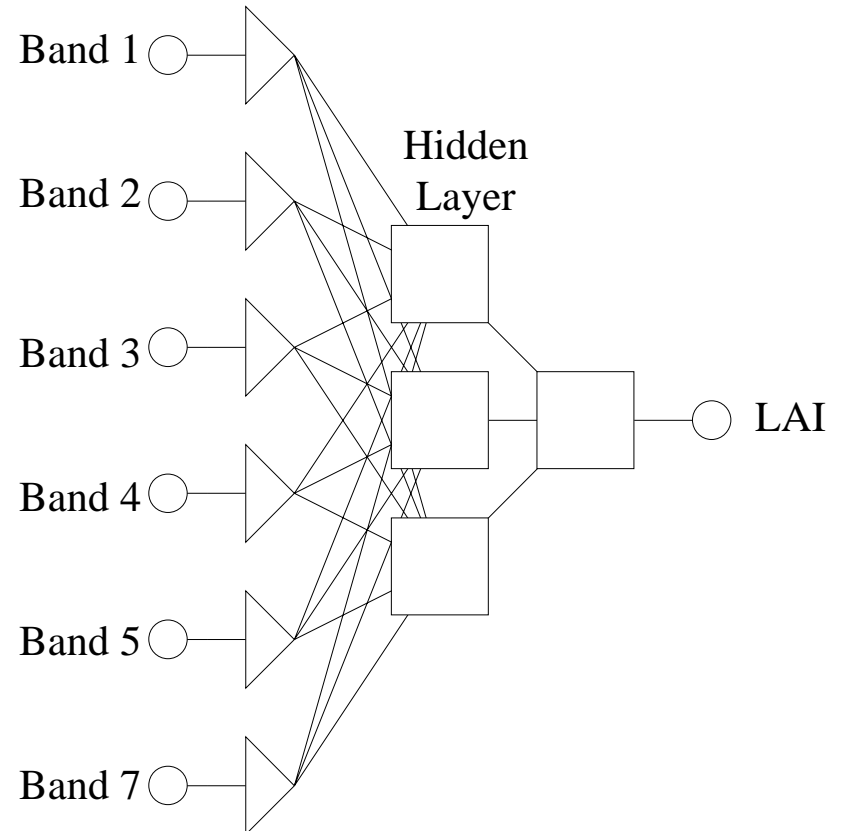
# Remote Urban LAI Estimation

- Good indicator of urban canopy
- Input into local, regional, and global models
- Terre Haute, Indiana
  - 145 random points
- Accuracy?

# ANN Method

- Artificial Neural Network
  - Backpropagation

Landsat TM Brightness Values



# Other Remote LAI Estimation Methods

- Parametric
- Regression
  - Vegetation indices
  - All band
  - Band ratios

# Terre Haute, Indiana, USA



# Terre Haute

- Population ~ 65,000
- Wabash River and Interstate 70
- County seat of Vigo County



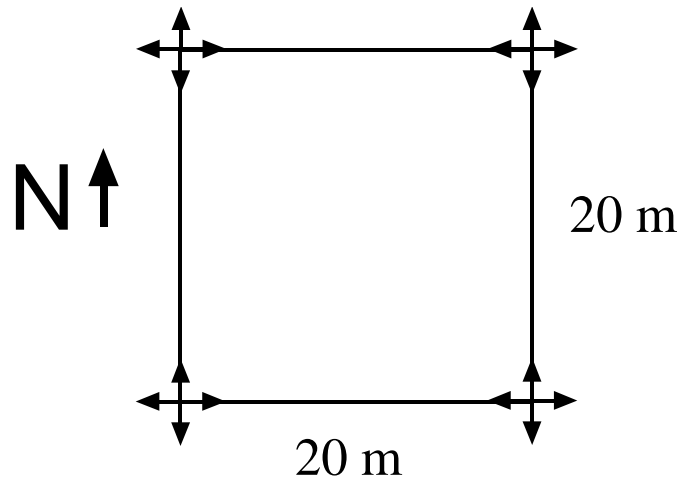
# Terre Haute's Urban Forest

- Fairly strict ordinance
- Rotating tree board
- City forester



# Sampling Method

- 20 meter quadrats
- GPS point



- Ceptometer



$$LAI = \frac{-\ln \frac{Q_1}{Q_2}}{k}$$

# Methods

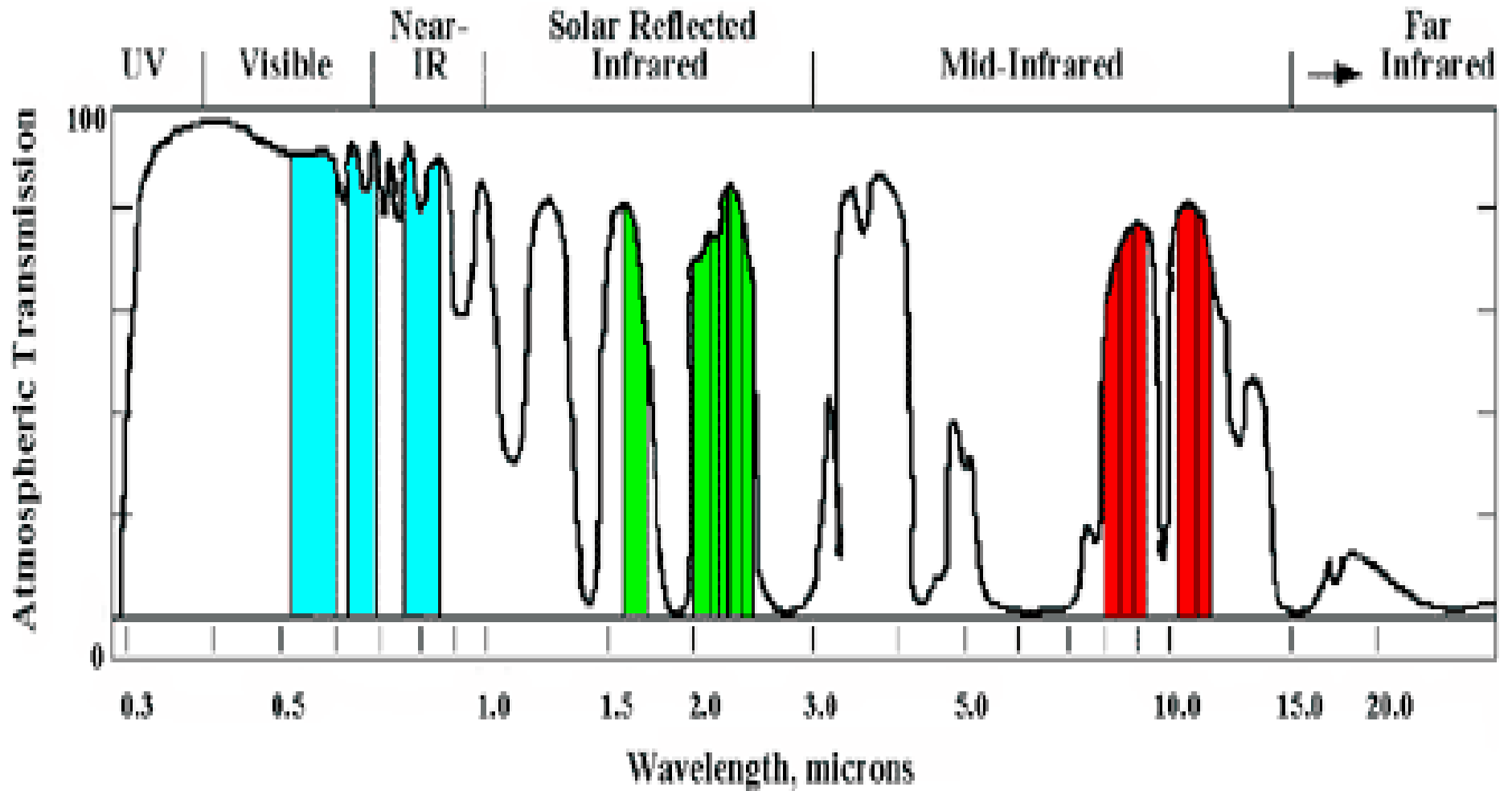
- Regression
  - Multiple inputs and ratios
- Artificial Neural Network
  - Multiple inputs and ratios



# ASTER

| Subsystem | Band No. | Spectral Range    | Spatial Resolution |
|-----------|----------|-------------------|--------------------|
|           |          | ( $\mu\text{m}$ ) |                    |
| VNIR      | 1        | 0.52 - 0.60       | 15 m               |
|           | 2        | 0.63 - 0.69       |                    |
|           | 3N       | 0.78 - 0.86       |                    |
|           | 3B       | 0.78 - 0.86       |                    |
| SWIR      | 4        | 1.600 - 1.700     | 30 m               |
|           | 5        | 2.145 - 2.185     |                    |
|           | 6        | 2.185 - 2.225     |                    |
|           | 7        | 2.235 - 2.285     |                    |
|           | 8        | 2.295 - 2.365     |                    |
|           | 9        | 2.360 - 2.430     |                    |
| TIR       | 10       | 8.125 - 8.475     | 90 m               |
|           | 11       | 8.475 - 8.825     |                    |
|           | 12       | 8.925 - 9.275     |                    |
|           | 13       | 10.25 - 10.95     |                    |
|           | 14       | 10.95 - 11.65     |                    |

# ASTER Spectral Bands



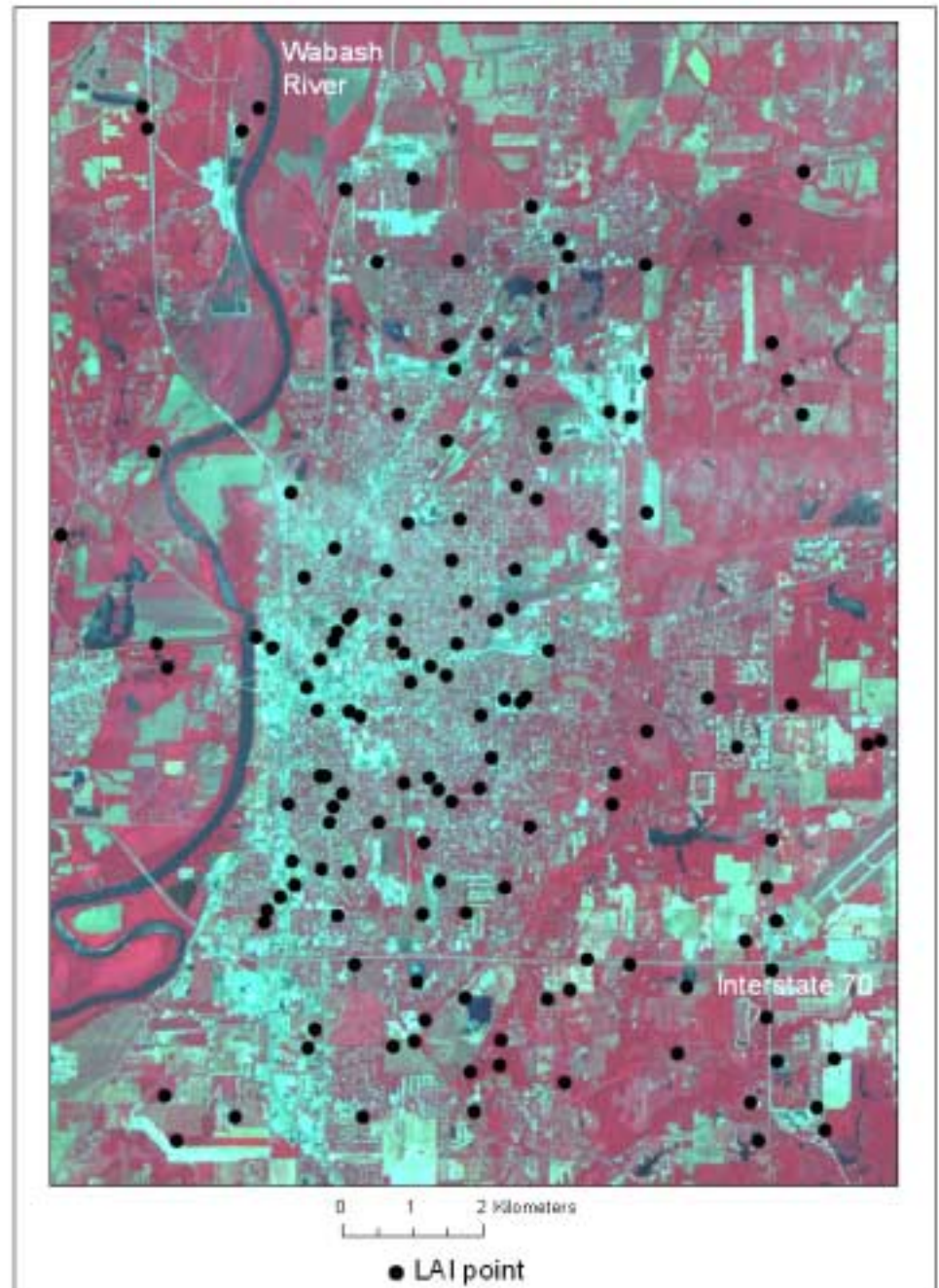
VNIR  
15 m

SWIR  
30 m

TIR  
90 m

# Leaf Area Index

- 145 random points
  - Regression
  - Artificial neural network



# Results

| <b>Method</b> | <b>Inputs</b>                | <b><i>R</i></b> | <b><i>SEE</i></b> |
|---------------|------------------------------|-----------------|-------------------|
| Regression    | <i>Green : Red</i>           | 0.60            | 1.54              |
| Regression    | <i>Green : Red, Infrared</i> | 0.62            | 1.51              |
| ANN           | <i>GREEN</i>                 | 0.69            | 1.39              |
| ANN           | <i>NDVI</i>                  | 0.68            | 1.39              |
| ANN           | <i>Red : Infrared</i>        | 0.69            | 1.39              |
| ANN           | <i>GREEN, Red : Infrared</i> | 0.71            | 1.35              |

# LAI Map

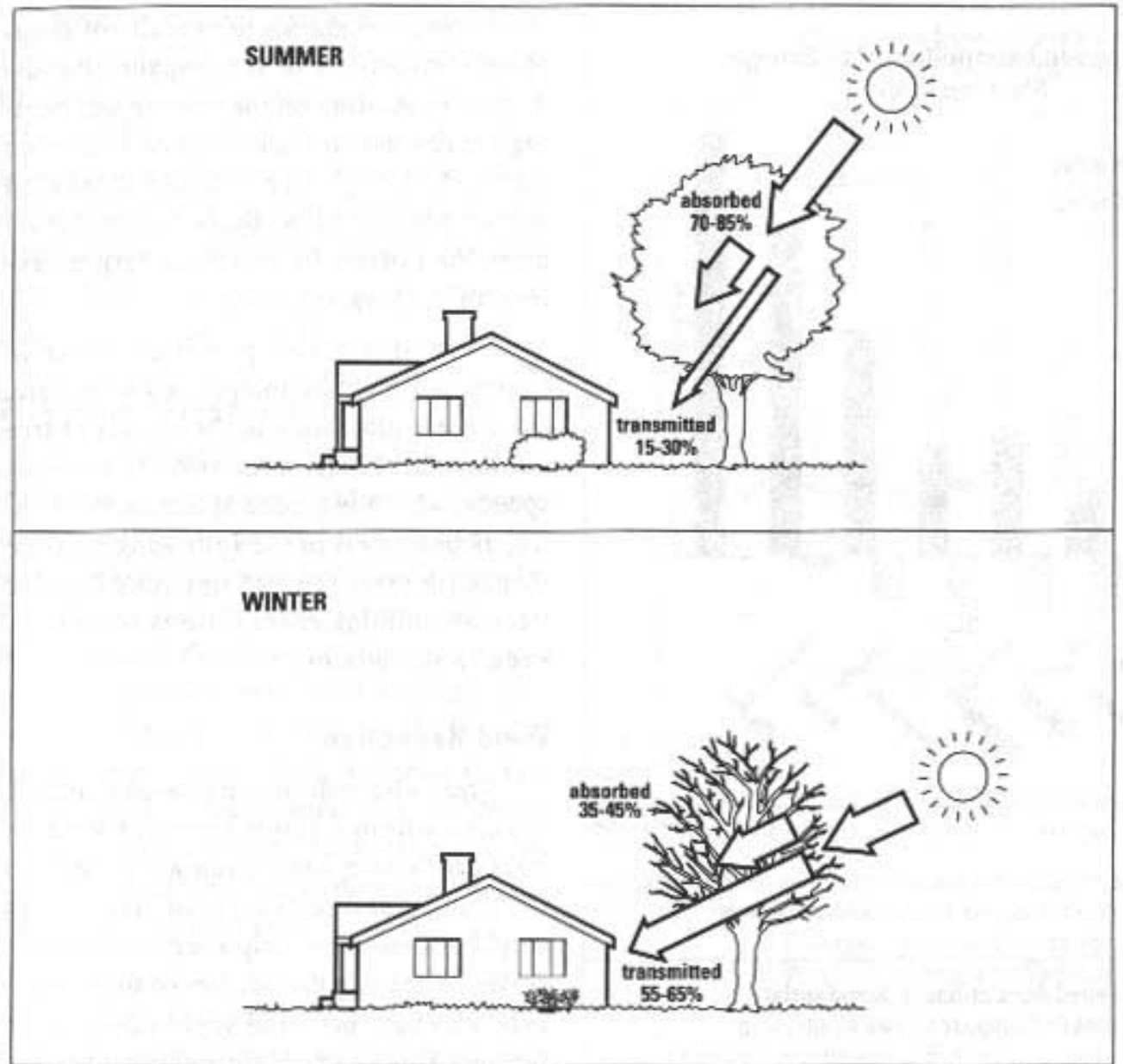
- Derived from ANN



# Questions

- Measurable difference in summer residential electricity use that can be attributed to the urban forest
- Relationship between urban leaf area and UHIE
- Equitable distribution of urban leaf area

# Energy Reduction



Source: Heister, 1996

Akbari et al., 1992, p. 29, fig 2-2.

# Terre Haute

- Relationship between LAI and summer energy consumption
- 300 random addresses
  - Issues of privacy
  - Cinergy PSI
  - 534 N. 6<sup>th</sup> Street



# Standardize

- Residential meters are generally read once each month, but not all on the same day.
  - Some residences appeared to have been missed one month with a make-up the next month
  - Some residences appeared to have been turned off for a month or more during the study time
- Kilowatt-hours/day.

# Correlation

- Slight correlation between LAI and household energy consumption
  - For every one unit increase in LAI, KwHr per day usage decreases by 4.17368



# Accuracy

- Summer time energy usage could depend on
  - Thermostat
  - Windows
  - Window units or central air
  - # of occupants
  - Insulation
  - Television

# Significance


- Money!
- ~ 120 cooling days
- \$0.075 / KwHr
  - One unit increase in LAI
    - \$37.53 season savings
  - Four unit increase in LAI
    - \$150.12

# Urban Heat Island

- Urban temperatures
- LAI

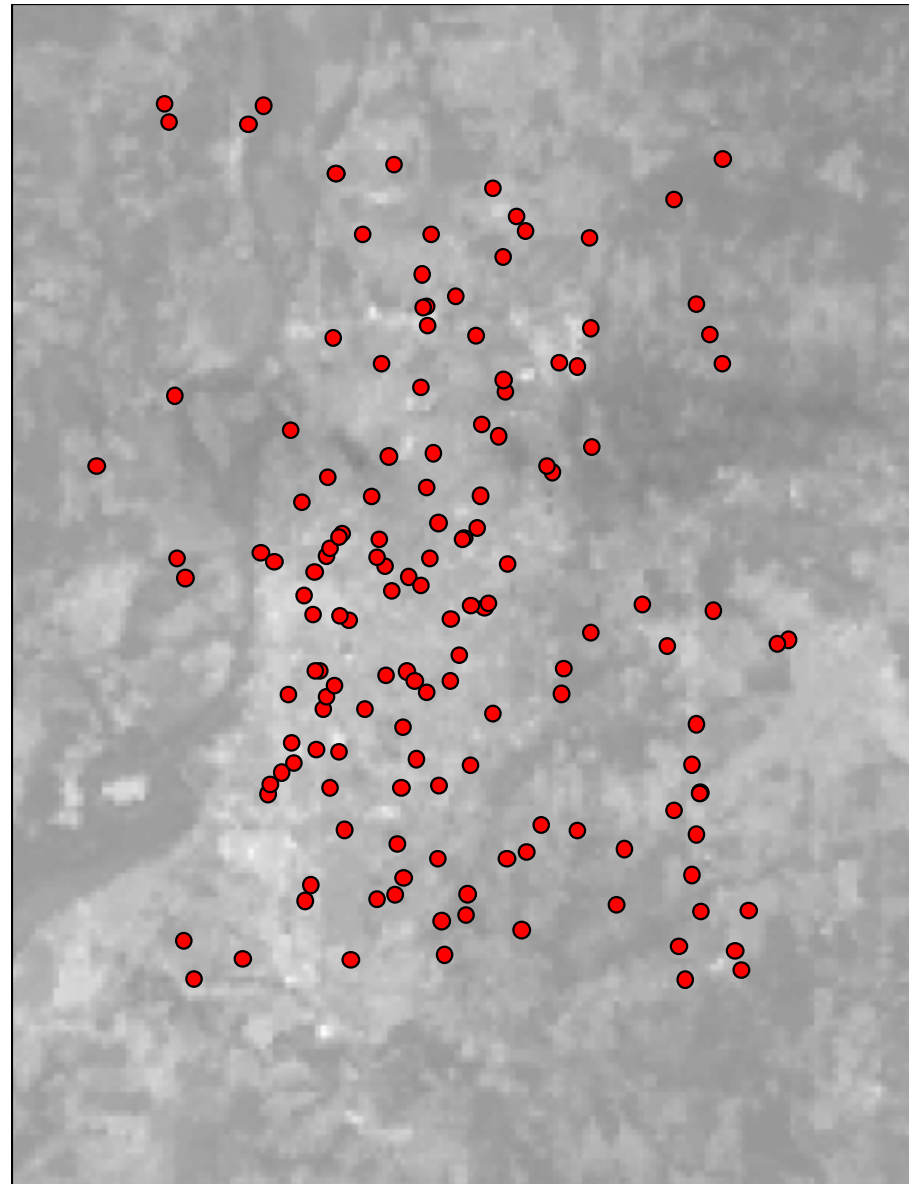


**Celsius**  High : 55.75  
Low : 0.0

0 1 2 4 Kilometers  


# Urban Temperature and LAI

- Same LAI dataset
  - 145 points
- Relationship with urban temperature



**Celsius**  High : 55.75  
Low : 0.0

0 1 2 4 Kilometers  


# Urban Temperature and LAI

- Explains 19.3% of variance
  - $R^2 = 19.3\%$
- Celsius =  $33.0 - 1.17 * LAI$
- Cooler temperatures as leaf area increases

# Equitable Distribution of UF

- Environmental justice
- Urban quality of life
- Distribution?
- 250 points
- ANN-derived urban LAI

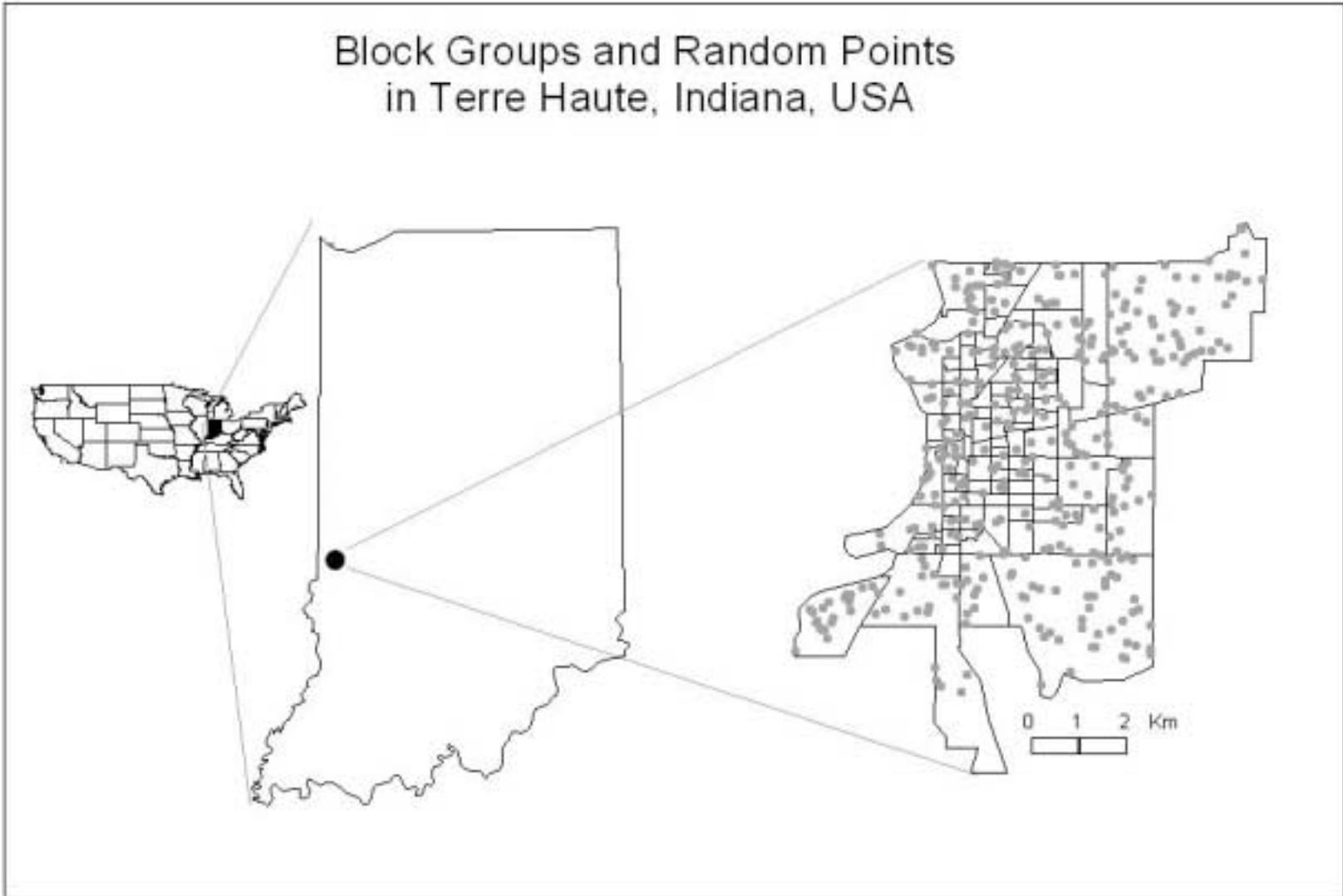


# Methods

- Census block groups
  - 2000
- Socio-economic patterns
  - Median income
  - Median housing value
  - Population density
  - % African American
  - Owner-occupied

# Random Points

Block Groups and Random Points  
in Terre Haute, Indiana, USA



# Results

|                     | LAI          | Home Value   | Median Income | Density      | Percent AA   |
|---------------------|--------------|--------------|---------------|--------------|--------------|
| LAI                 | 1            |              |               |              |              |
| Home Value          | 0.321        | 1            |               |              |              |
| <i>significance</i> | <i>0.000</i> | <i>0.000</i> |               |              |              |
| Median Income       | 0.401        | 0.710        | 1             |              |              |
| <i>significance</i> | <i>0.000</i> | <i>0.000</i> | <i>0.000</i>  |              |              |
| Density             | -0.325       | -0.465       | -0.489        | 1            |              |
| <i>significance</i> | <i>0.000</i> | <i>0.000</i> | <i>0.000</i>  | <i>0.000</i> |              |
| Percent AA          | -0.064       | -0.232       | -0.232        | 0.392        | 1            |
| <i>significance</i> | <i>0.225</i> | <i>0.000</i> | <i>0.000</i>  | <i>0.000</i> | <i>0.000</i> |
| Owner Occ.          | -0.156       | -0.386       | -0.326        | 0.078        | 0.324        |
| <i>significance</i> | <i>0.004</i> | <i>0.000</i> | <i>0.000</i>  | <i>0.137</i> | <i>0.000</i> |

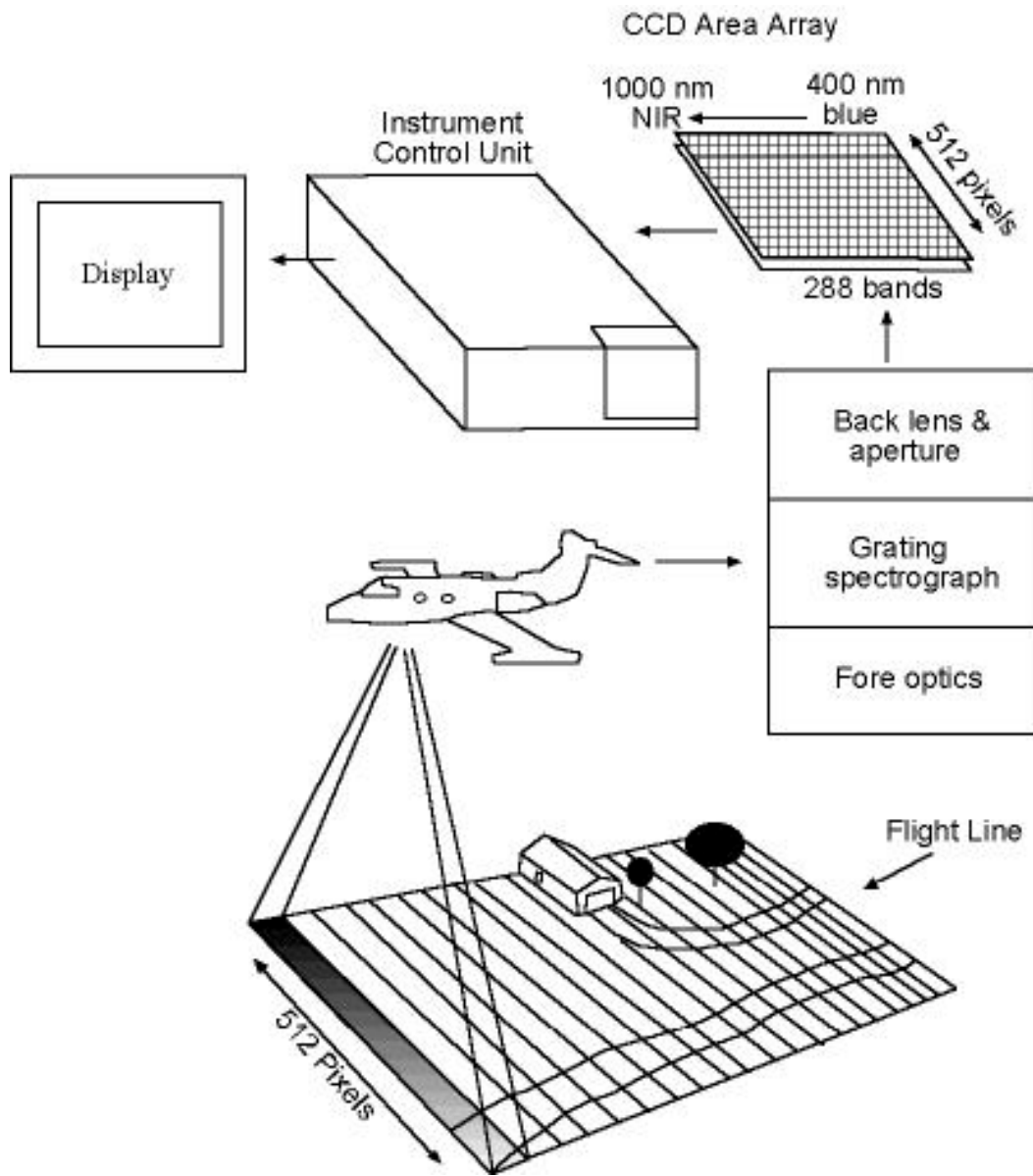
# New Urban Image Data

- AISA+ Hyperspectral Sensor
  - Recently awarded to ISU from National Science Foundation
  - 400 – 900 nm
    - Spectral bandwidth 2.3 nm
- Portable
- Bands are fully programmable

# AISA+ Hyperspectral Sensor



# AISA+ Hyperspectral Sensor





**RESEARCH PARTNERS**

**University of South Carolina**  
**Center for GIS & Remote Sensing**  
**John Jensen, Co-Director**

**Indiana University**  
**Anthropological Center for Training**  
**Emilio Moran, Director**

NASA/LBA Ecology      Urban Mapping      UTPA/URDA

**Current External Collaborations**

Wabash View      Fire Ecology Research Group      Alternative Policy Assessment Regimes

**Planned/Future Collaborations using hyperspectral data as extensions of current research clusters**

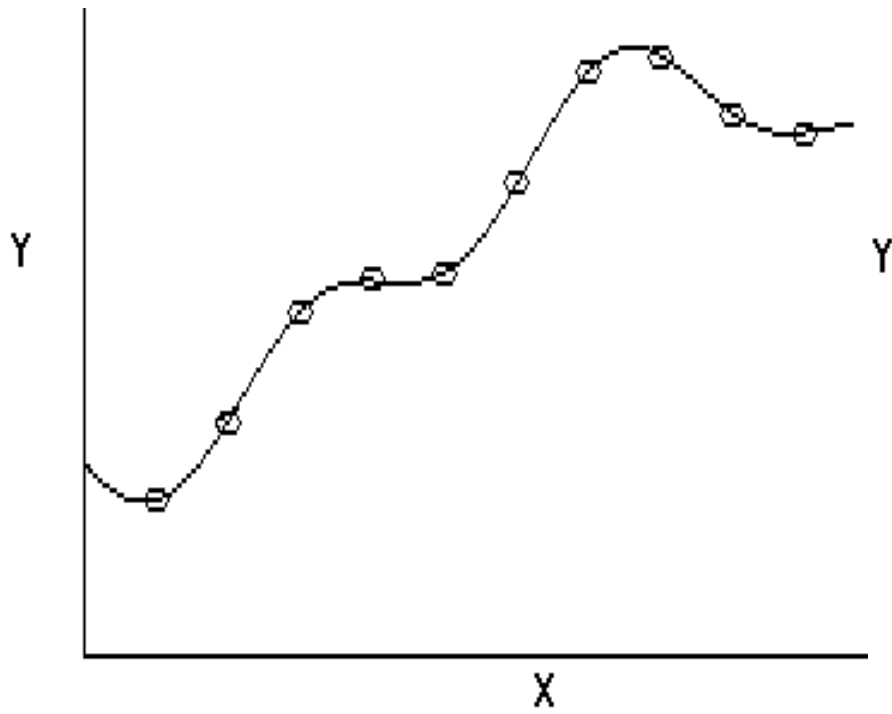
# Conclusions

- Remote sensing
- Different kinds of studies
- ASTER
- Hyperspectral
- Qualitative research
- Reliable field data



# How many neurons?

Overfit



Goodfit

