7-County Metro Area Change Detection

By Tyler Dardis

### What's the 7-**County Metro** Area? Anoka T Washington Hennepin Ramsey 1 -27 -Carver--AF Dakota Scott

### • Counties that contain, and surround the Twin Cities

- Today, comprise approx. 3.5 million people, over half the state's population
- Includes:
  - Anoka
  - Hennepin (Minneapolis)
  - Ramsey (St. Paul)
  - Washington
  - Dakota
  - Scott
  - Carver

# What's the Question?

• Answer the question of when and where the 7-Country Metro Area has expanded between 1975 and 2015.

# Why?

• Urban Growth can have many effects on the environment, destroying ecosystems, and much more

• To answer the question if the metro area is doing urban revitalization (urban renovation, keeping within limits), or rampantly expanding and sprawling away from the core.

### How?

• Take remotely sensed imagery, and **classify** it into Urban, and Non-Urban areas. You can then see where change has occurred through change detection mapping.

• You can also get area data to see change in numbers.

### Data Used





All Landsat Data from USGS's <u>GloVis</u>
1975 – Landsat 1 (July 29<sup>th</sup>)
1985 – Landsat 5 (April 28<sup>th</sup>)
1995 – Landsat 5 (July 13<sup>th</sup>)
2005 – Landsat 5 (June 22<sup>nd</sup>)
2015 – Landsat 8 (March 14<sup>th</sup>)

MN Statewide Country Shapefile from <u>MNDOT GIS Data</u>
 Used to create 7-country metro AOI

Need to Pre-Process the Data Start with Stacking Bands into one image
 Under Raster -> Spectral -> Layer Stack

- Stacked bands 4-7 for LS1
- Stacked bands 1-7 for LS5 and LS8
  Saved all as one single .img



# Mosaicking

• 1985-2005 didn't have just one Landsat cover the 7-county area

- Needed 2 Landsat images, and mosaic them together
- Once the both images are stacked, we simply open MosaicPro
  - Raster -> Mosaic -> MosaicPro
- Added both images
- Changed Seamline to Weighted (blends together)
- Ran the Process
  - Saved both images, mosaiced together, as one .img file



# Clipping it Down

- Landsat Imagery covers a large area; only need 7-county area
  - Imported MNDOT's statewide county vector shapefile
  - Selected the 7 counties desired
  - With the counties selected, under the Vector Drawing tab, click Paste from Selected Object
    - This makes an AOI over the selected vector

# Clipping it Down Continued

- With the 7-county AOI, we can begin the clipping process (called Subset)
  - Raster -> Subset & Chip -> Subset
- Selected input as the Landsat image, or mosaiced image
- Output as '\_clipped'
- Click AOI
  - Select 7-county AOI







# Haze Reduction if Needed

• Haze reduction can help remove distortions in imagery due to fog, cloudy haze.

- In my case, major areas were misclassified
- Ended up finding new Landsat imagery for the most accuracy

# Here's Why...

Northwest corner of Anoka county misclassified to suburban due to thin haze.



# Classification Scheme Used

Supervised Classification

- Classified into Water, Urban, Suburban, Cropland/Non-Urban, and Forest area
  - Values started at 1, ended at 5
- Used **8** training polygons per class, per image

Urban + Suburban Acreage =





Urban + Suburban Acreage =



Urban + Suburban Acreage =





Urban + Suburban Acreage =



Urban + Suburban Acreage =





# Issues with Classification

Cloud interference

- Misclassification on small scale, causing figure changes
  Boundary of urban area still grows
- Problems with images themselves (brightness, clarity)







 Accuracy Assessment is a nice benchmark to see how well you classified your data

• Raster -> Supervised -> Accuracy Assessment

• Used 50 random points

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### ACCURACY TOTALS

Class Name	Reference Totals	Classified Totals	Number Correct	Producers Accuracy	Users Accuracy
Unclassified Vater Urban Suburban Cropland, Non-U Forest	0 1 2 13 32 2	0 (2) 34 3	0 1 1 29 29 2	 100.00% 50.00% 69.23% 90.63% 100.00%	 100.00% 50.00% 90.00% 85.29% 66.67%
Totals	50	50	42		
Overall Classifi	cation Accuracy	= 84.00%			

----- End of Accuracy Totals -----

### KAPPA (K^) STATISTICS

Overall Kappa Statistics = 0.6853

Conditional Kappa for each Category.

Class Name	Kappa
Unclassified Water Urban Suburban Cropland, Non-Urban	0.0000 1.0000 0.4792 0.8649 0.5915 0.6528

----- End of Kappa Statistics -----

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#### ACCURACY TOTALS

Class Name	Reference Totals	Classified Totals	Number Correct	Producers Accuracy	Users Accuracy
Unclassified Water Urban Suburban Cropland, Non-U Forest	0 2 3 3 32 10	 0 2 3 2 30 13	0 2 3 2 29 10	 100.00% 100.00% 66.67% 90.63% 100.00%	 100.00% 100.00% 100.00% 96.67% 76.92%
Totals	50	50	46		
Overall Classifi	cation Accuracy	· = 92.00%			

----- End of Accuracy Totals -----

### KAPPA (K^) STATISTICS

Overall Kappa Statistics = 0.8562

Conditional Kappa for each Category.

Class Name	Kappa		
Unclassified	0.0000		
Vater	1.0000		
Urban	1.0000		
Suburban	1.0000		
Cropland, Non-Urban	0.9074		
Forest	0.7115		

----- End of Kappa Statistics -----

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### ACCURACY TOTALS

Class	Reference	Classified	Number	Producers	Users
Name	Totals	Totals	Correct	Accuracy	Accuracy
Unclassified Water Urban Suburban Cropland, Non-U Forest Totals	0 2 1 13 28 6 50	 0 1 16 29 2 50	0 2 1 13 26 2 44	 100.00% 100.00% 100.00% 92.86% 33.33%	 100.00% 100.00% 81.25% 89.66% 100.00%

Overall Classification Accuracy = 88.00%

----- End of Accuracy Totals -----

#### KAPPA (K^) STATISTICS

Overall Kappa Statistics = 0.7949

Conditional Kappa for each Category.

Class Name	Kappa
Unclassified	0.0000
Water	1.0000
Urban	1.0000
Suburban	0.7466
Cropland, Non-Urban	0.7649
Forest	1.0000

----- End of Kappa Statistics -----

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### ACCURACY TOTALS

Class	Reference	Classified	Number	Producers	Users
Name	Totals	Totals	Correct	Accuracy	Accuracy
Unclassified	0	0	0		
Water	2	2	2	100.00%	100.00%
Urban	4	4	4	100.00%	100.00%
Suburban	15	11	11	73.33%	100.00%
Cropland, Non-U	25	30	25	100.00%	83.33%
Forest	4	3	3	75.00%	100.00%
Totals	50	50	45		

Overall Classification Accuracy = 90.00%

----- End of Accuracy Totals -----

#### KAPPA (K^) STATISTICS

Overall Kappa Statistics = 0.8390

Conditional Kappa for each Category.

Class Name	Kappa		
Unclassified	0.0000		
Vater	1.0000		
Urban	1.0000		
Suburban	1.0000		
Cropland, Non-Urban	0.6667		
Forest	1.0000		

----- End of Kappa Statistics -----

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### ACCURACY TOTALS

Class Name	Reference Totals	Classified Totals	Number Correct	Producers Accuracy	Users Accuracy
Unclassified	0	0	0		
Water	5	3	3	60.00%	100.00%
Urban	0	2	0		
Suburban	11	11	11	100.00%	100.00%
Cropland, Non-R	30	23	22	73.33%	95.65%
Forest	4	11	3	75.00%	27.27%
Totals	50	50	39		

Overall Classification Accuracy = 78.00%

----- End of Accuracy Totals -----

### KAPPA (K^) STATISTICS

Overall Kappa Statistics = 0.6626

Conditional Kappa for each Category.

Class Name	Kappa
Unclassified	0.0000
Water	1.0000
Urban	0.0000
Suburban	1.0000
Cropland, Non-Rural	0.8913
Forest	0.2095

----- End of Kappa Statistics -----

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Issues with Accuracy Assessment

• Many of the random points were in the non-urban area

- Very few in urban, suburban areas
  - If chosen one wrong, can make accuracy 50% or less

# Results

Final Change Detection Map from 1975-2015
 Answers the question of where change happened

• Trend of Urban Growth in comparison with Population Growth

• Answers the question of **when** change happened

• Is it speeding up, slowing down?

# **Change Detection**



From **1975-2015** 

Total Additional Urban Acreage =



Concentrations on fringe, and dispersion as you go further out



# Change Detection Table

Row	Histogram	1975_sup_class value	2015_sup_class value	Color	Opacity	Area
0	3243836	0	0		1	721413
1	277301	1	1		1	61670.4
2	27815	1	2		1	6185.92
3	498	1	3		1	110.753
4	1267	1	4		1	281.774
5	1110	1	5		1	246.858
6	41558	2	1		1	9242.29
7	174695	2	2		1	38851.3
8	97945	2	3		1	21782.5
9	47096	2	4		1	10473.9
10	5237	2	5		1	1164.68
11	14256	3	1		1	3170.46
12	201137	3	2		1	44731.9
13	869905	3	3		1	193463
14	468401	3	4		1	104170
15	100379	3	5		1	22323.8
16	10190	4	1		1	2266.21
17	273167	4	2		1	60751
18	1263607	4	3		1	281020
19	2909925	4	4		1	647153
20	574979	4	5		1	127872
21	2765	5	1		1	614.922
22	50744	5	2		1	11285.2
23	212706	5			1	47304.8
24	362031	5	4		1	80513.9
25	576698	5	5		1	128255

# Observations

New Suburbs are much more spread out, not cramped
Could have also caused issues with classifications





### Acerage Growth in 7-County Metro Area



### Population Growth in 7-County Metro Area



# Results

- According to three points of data:
  - Urban Growth is steady to slowing
  - Could indicate slowing of urban sprawl
    - Increase in Urban Revitalization
  - Restrictions of expansion, Urban Growth Boundaries (UGB)
    - Needs more data
  - Supported by continued increase in population in recent years

 Project opens the door for continued and more detailed research into trends of the metro's growth in relation to area, and population

## Improvements

- More accurate and defined classification
- Use of higher resolution imagery for detailed, accurate classifications
  - Use of Feature Analyst
- Research into UGBs in Minnesota Metro
- Greater understanding of both the area, and image interpretation