

Dynamic Neighborhood Taxonomy

A Project of

Living Cities

Presentation by

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RW Ventures, LLC



Agenda

DNT: Project Overview

Measuring Change: the RSI

**Analytic Applications: Where to Invest; Pace,
Degree of Change; Role of Region; Drivers**

Evolution: Discovering Patterns of Change

Developing Tools: from Diagnostics to Investment



About Living Cities



“A partnership of financial institutions, national foundations and federal government agencies that invest capital, time and organizational leadership to advance America’s urban neighborhoods.”

Living Cities Partners:

AXA Community Investment Program
Bank of America
The Annie E. Casey Foundation
J.P. Morgan Chase & Company
Deutsche Bank
Fannie Mae Foundation
Ford Foundation
Bill & Melinda Gates Foundation

Robert Wood Johnson Foundation
John S. and James L. Knight Foundation
John D. and Catherine T. MacArthur Foundation
The McKnight Foundation
MetLife, Inc.
Prudential Financial
The Rockefeller Foundation
United States Department of Housing & Urban Development



Partners and Advisors



... And Over 70 Advisors including Practitioners, Researchers, Funders, Civic Leaders and Government Officials



We Know Where We Want to Go...

Common Goal:



BUILDING HEALTHIER COMMUNITIES

The Challenge: Scarce Resources, Many Options

- Community-Based Organizations: select interventions, identify assets and attract investment
- Governments: tailor policy and interventions
- Businesses: identify untapped neighborhood markets
- Foundations: evaluate interventions

**Need for Relevant, Timely and
Accessible Information Resources**



Information Resources



Data

**Increasingly available,
but more progress to be made**

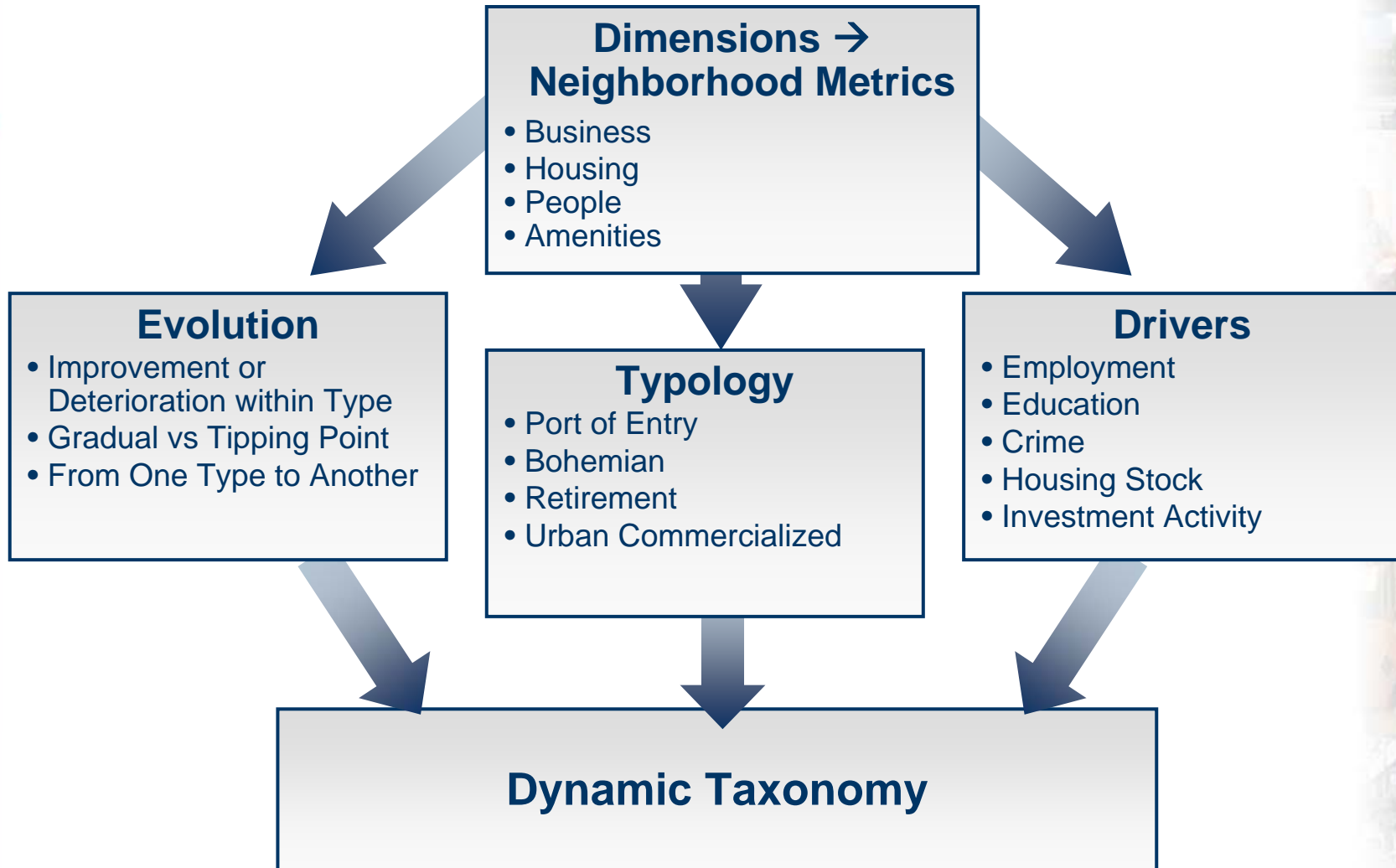
Knowledge

**Gap between
practitioners and academics:
need “Clinical Economics”
(Sachs)**

Tools

**Few decision systems for
neighborhood practitioners
and investors**

Comprehensive Neighborhood Taxonomy



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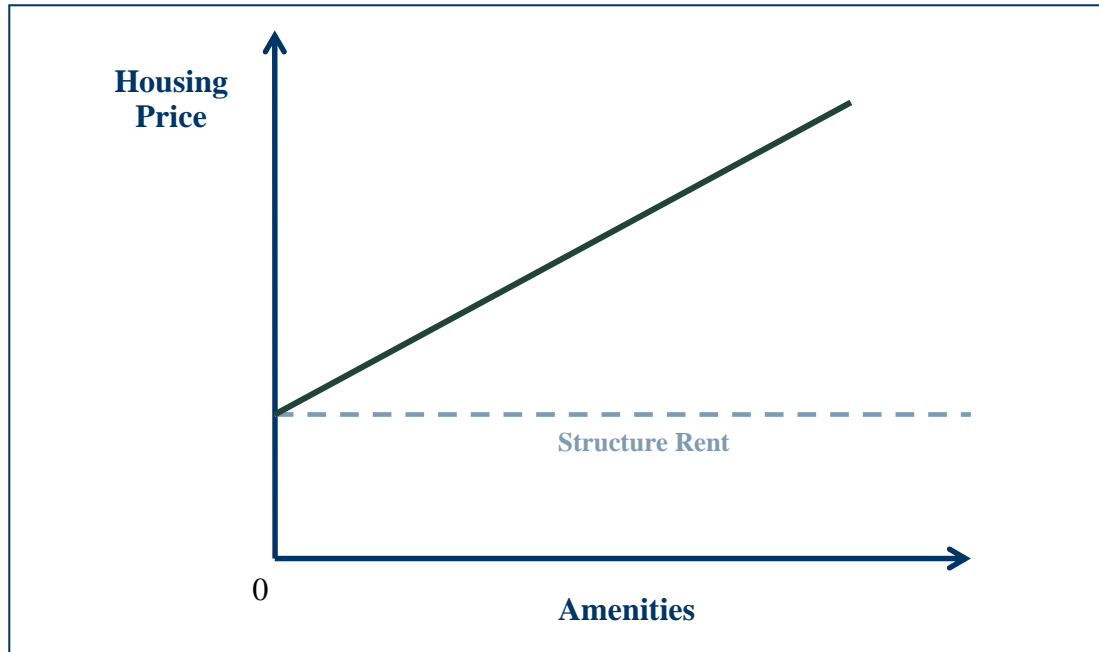
Analytic Applications: Where to Invest; Pace,
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Theoretical Framework



- Use Demand for Housing as Proxy for Neighborhood Health
- Look at Housing Values to Capture Neighborhood Amenities
- Look at Change in Quantity of Housing to Account for Supply Effects



The Challenge: Finding a Metric that Works

Issues:

- Measure change in prices controlling for change in quality of the housing stock
- Estimate at very small level of geography
- Track continuous change over time

Solutions:

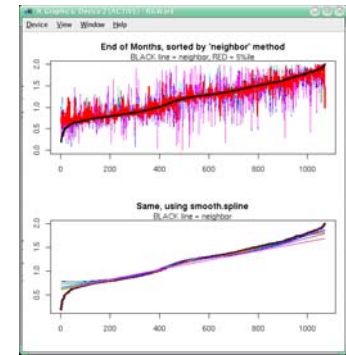
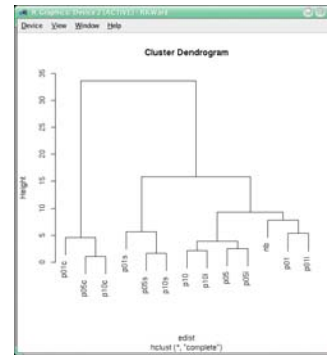
- Repeat Sales to Control for Changes in Neighborhood Housing Stock
- Spatial Smoothing: Locally Weighted Regression to account for “fluid” neighborhood boundaries and address sample size
- Temporal Smoothing: Fourier expansions to track change over time



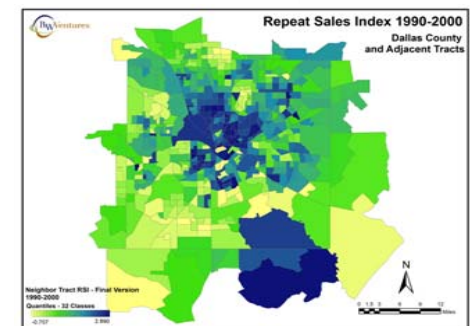
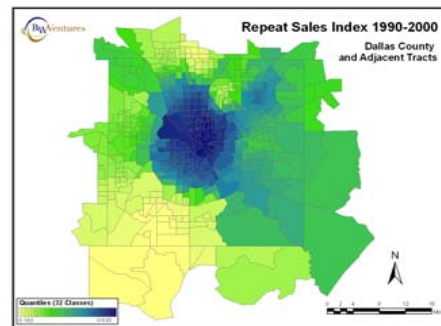
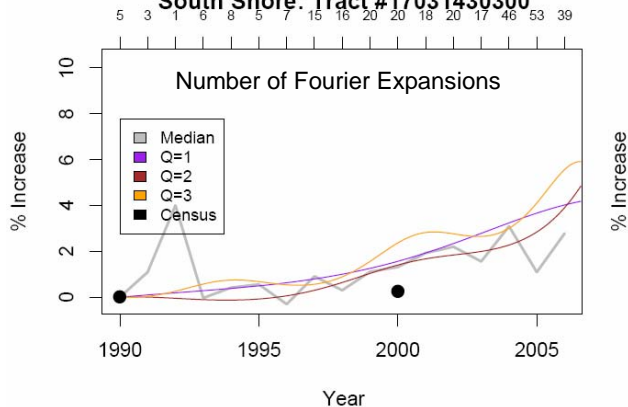
Developing the Index: Spatial and Temporal Smoothing

Correlations between different RSI Versions

	p01	p01i	p01s	p01c	p05	p05i	p05s	p05c	p10	p10i	p10s	p10c	nb
p01													
p01i	0.96												
p01s	0.82	0.89											
p01c	0.53	0.58	0.82										
p05	0.94	0.96	0.79	0.49									
p05i	0.94	0.97	0.82	0.52	0.99								
p05s	0.83	0.90	0.99	0.77	0.84	0.87							
p05c	0.53	0.59	0.82	1.00	0.50	0.53	0.79						
p10	0.92	0.94	0.76	0.47	0.99	0.99	0.82	0.49					
p10i	0.92	0.95	0.79	0.50	0.98	0.99	0.85	0.51	0.99				
p10s	0.84	0.90	0.97	0.75	0.85	0.88	1.00	0.77	0.83	0.86			
p10c	0.53	0.59	0.82	0.99	0.51	0.54	0.79	1.00	0.49	0.51	0.77		
nb	0.11	0.13	0.11	0.07	0.13	0.13	0.12	0.07	0.13	0.13	0.12	0.07	



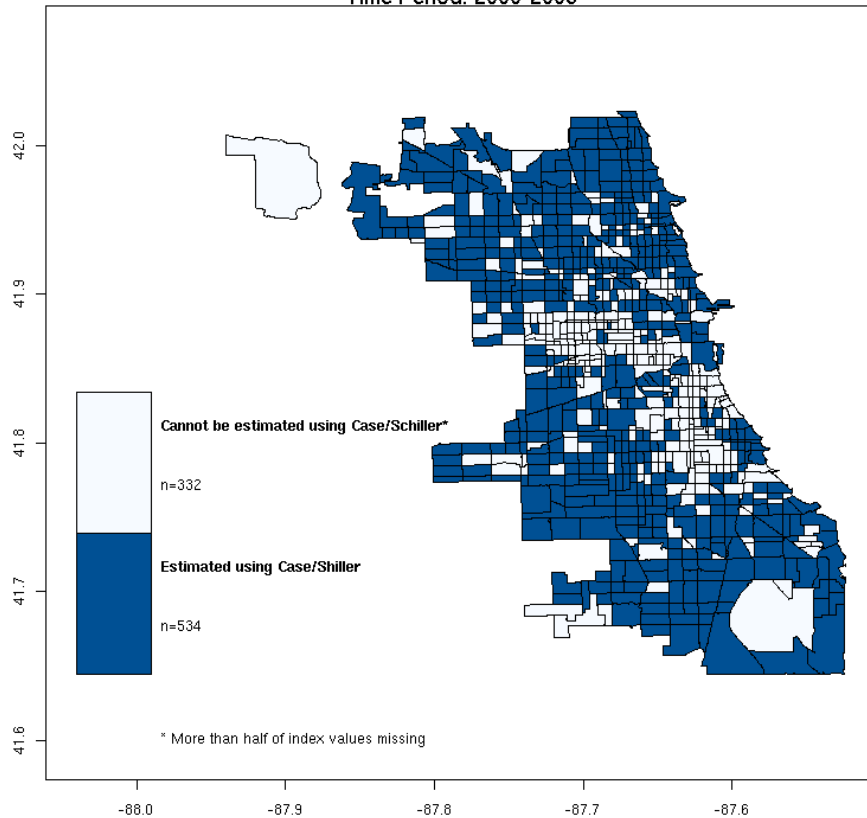
South Shore: Tract #17031430300



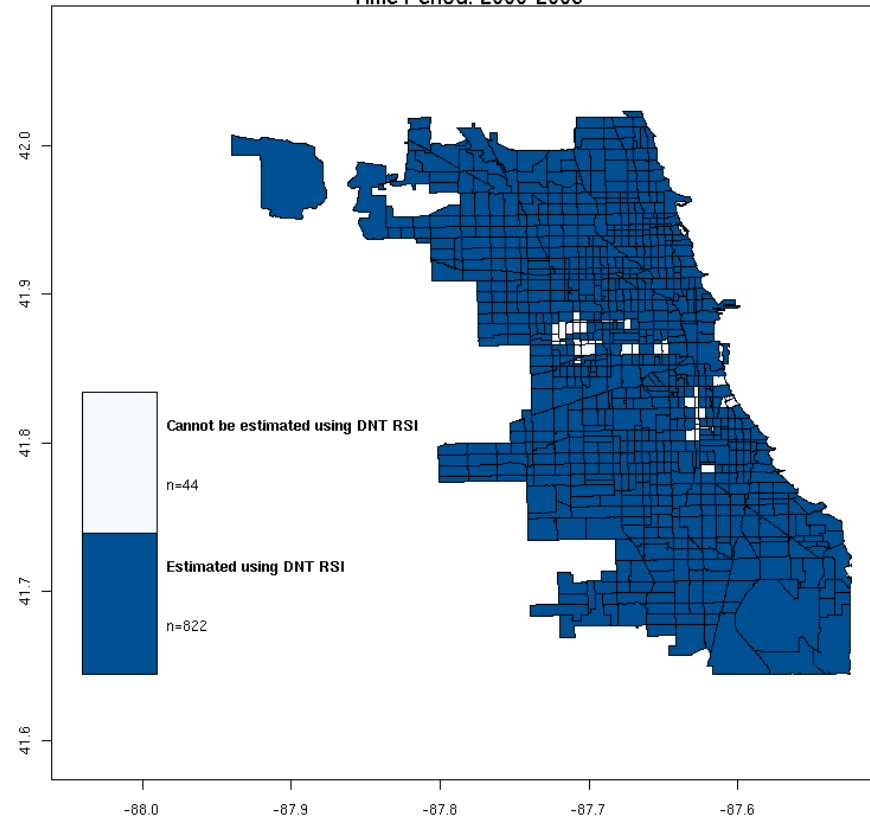
Optimizing sample size and fluid boundaries through extensive modeling and cross-validation procedures

Final Product: The DNT RSI

RSI Estimation Coverage Using Case/Shiller Method
Time Period: 2000-2006



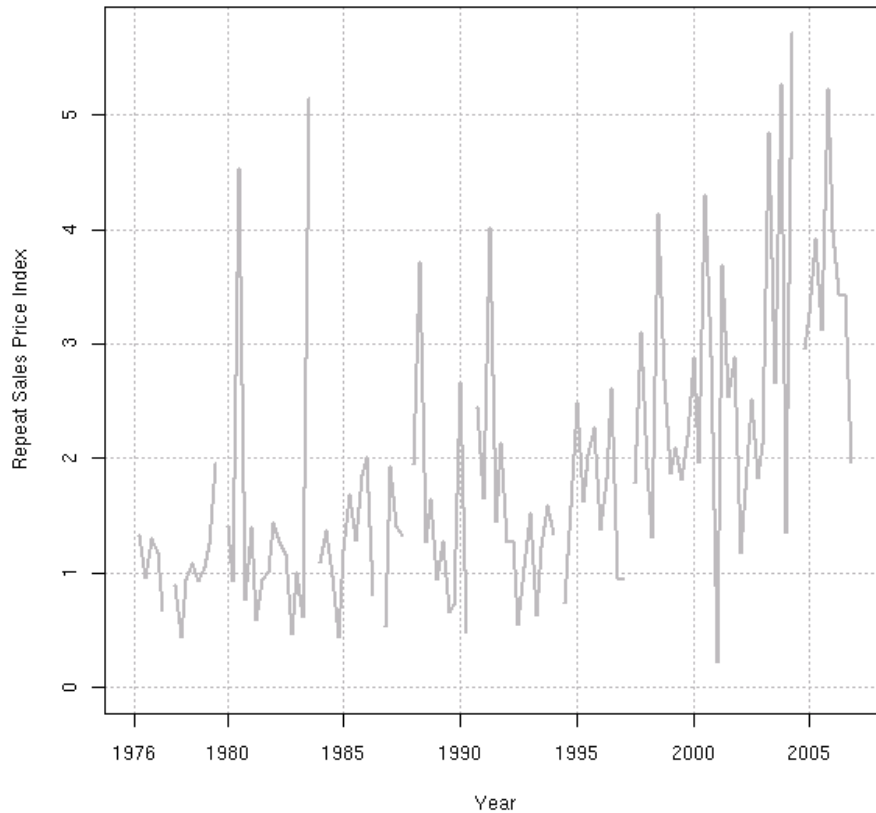
RSI Estimation Coverage Using DNT RSI Method
Time Period: 2000-2006



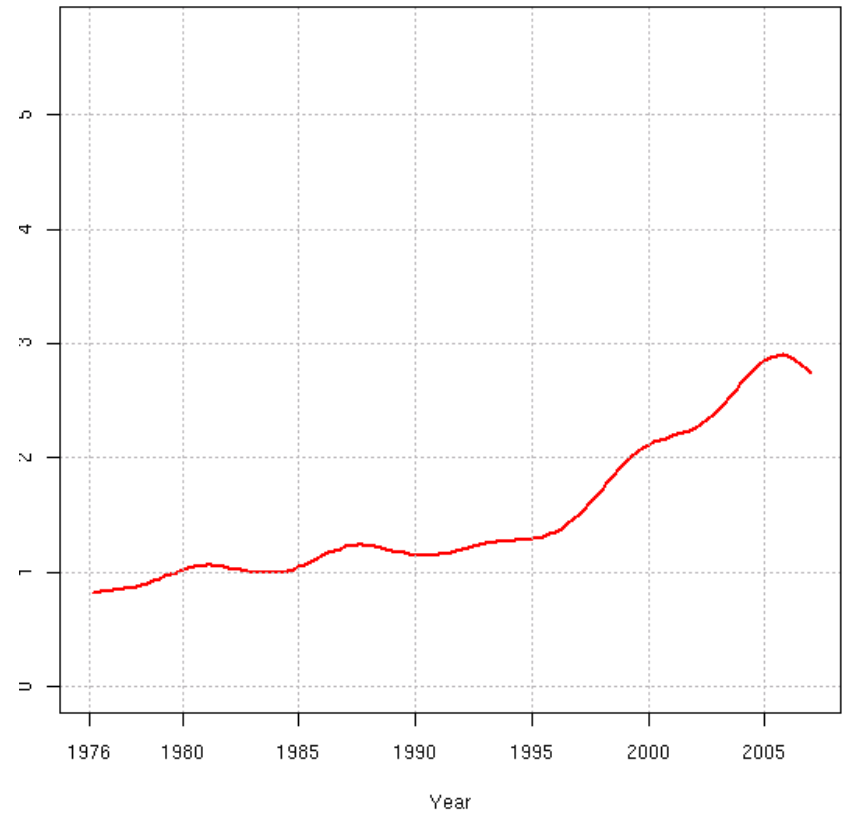
Unlike traditional repeat sales indices, the DNT RSI can be estimated for very small levels of geography

Final Product: The DNT RSI

Case/Shiller Index for Tract 39035115200 in Cleveland

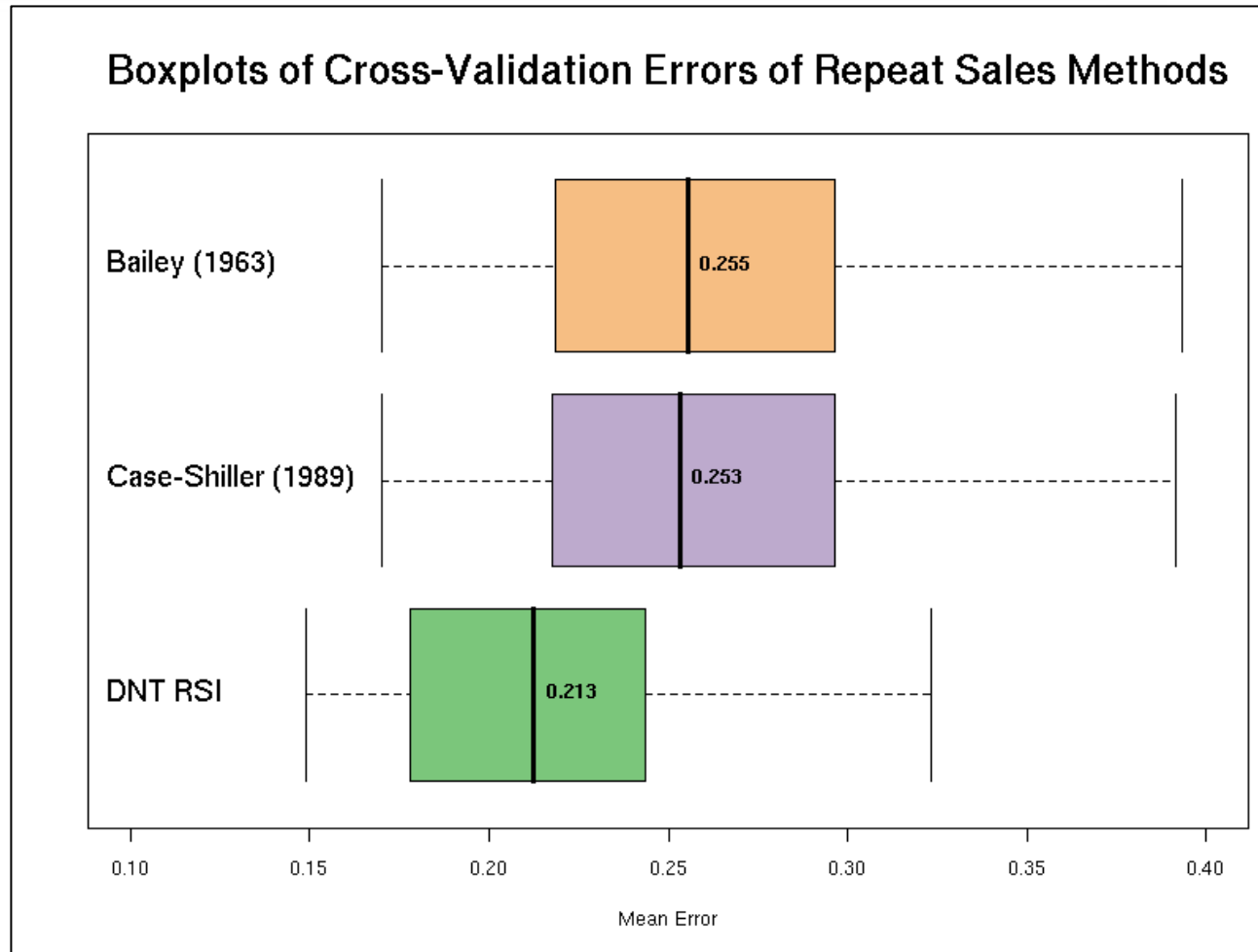


DNT Repeat Sales Index for Tract 39035115200 in Cleveland



Less volatile than traditional RSIs

Final Product: The DNT RSI



More robust than traditional repeat sales indices at the tract level

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DNT: Project Overview

Measuring Change: the RSI

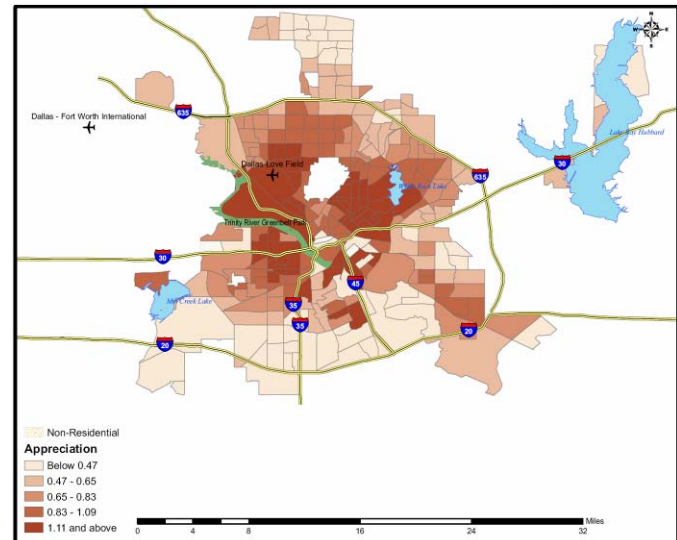
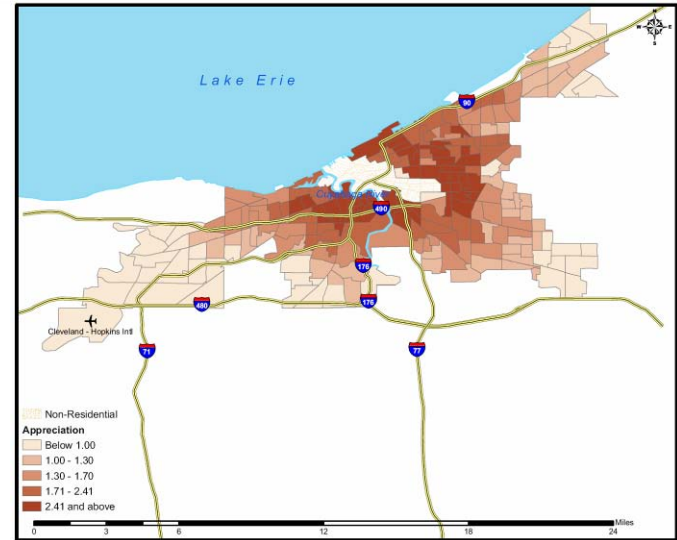
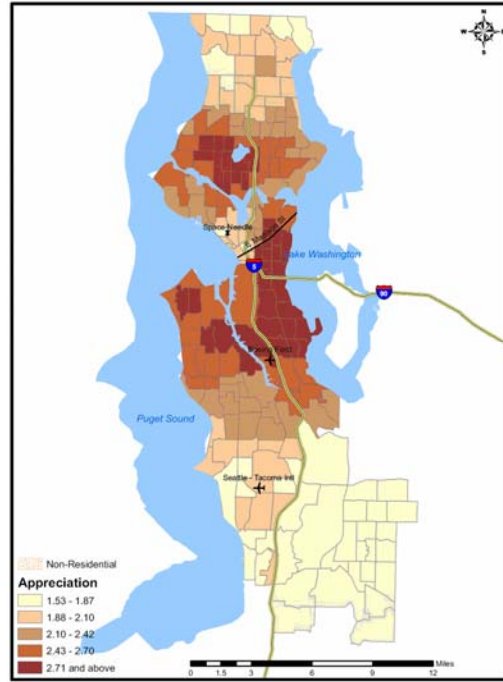
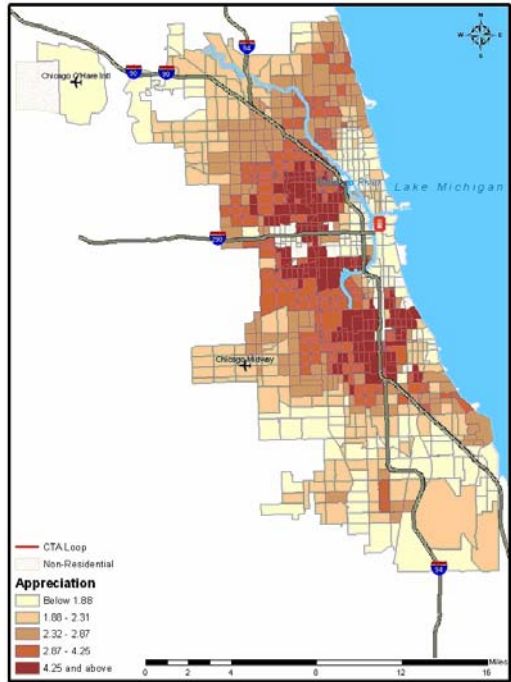
**Analytic Applications: Where to Invest; Pace,
Degree of Change; Role of Region; Drivers**

Evolution: Discovering Patterns of Change

Developing Tools: from Diagnostics to Investment

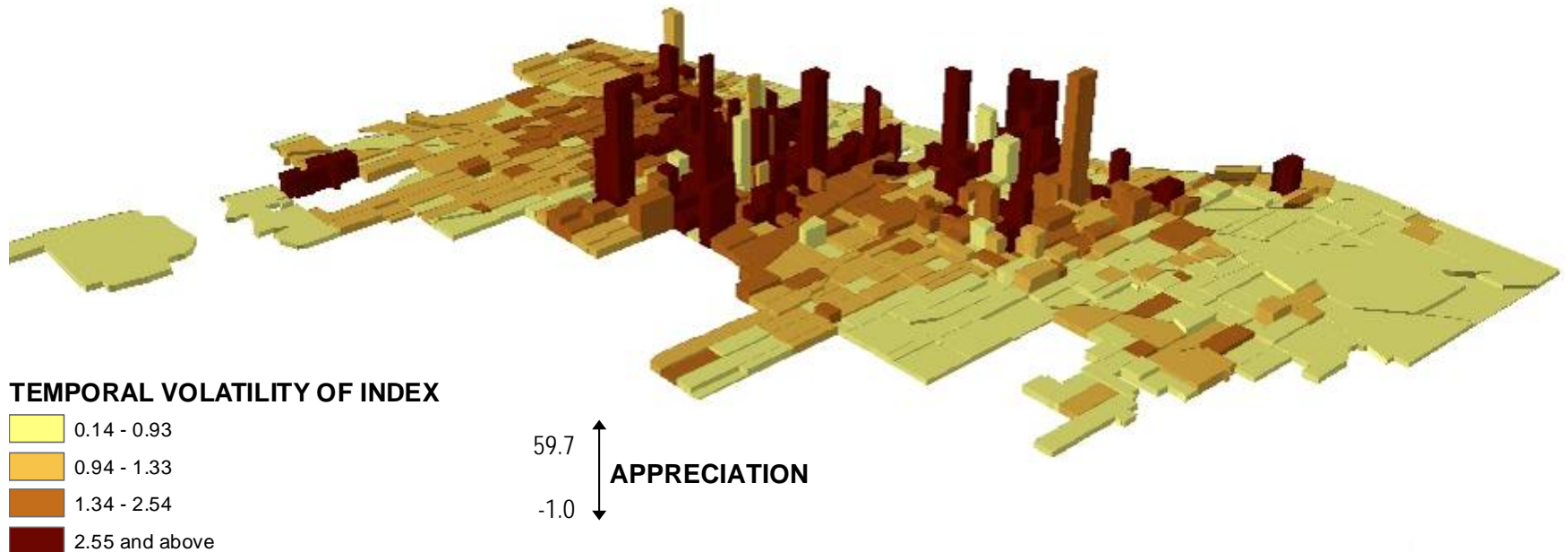


Change in Price: Poor Neighborhoods Present the Most Opportunities for Investment



Many of the poorest neighborhoods are the ones that grew the most, outperforming wealthier communities in each of the four sample cities

Partly Due to Lack of Information, These Areas Are Also the Most Volatile



By increasing the availability of information on these markets, we could reduce risk, increase market activity, and help stabilize these communities, further strengthening their performance.

Using the RSI to Develop New Knowledge

■ How Much and How Fast do Neighborhoods Change?

- Neighborhood change is a slow process: over 15 years, most neighborhoods don't change their position relative to other neighborhoods in the region.

(Methodology: Transition Matrices)

■ How Important Is the Region?

- Across cities, 35% of all neighborhood change is accounted for by regional trends.

(Methodology: Correlations and Regressions)

■ Do Neighborhoods “Converge”?

- Overall, neighborhoods tend to “catch up” with each other, but there are important exceptions

(Methodology: Sigma and Beta Convergence)



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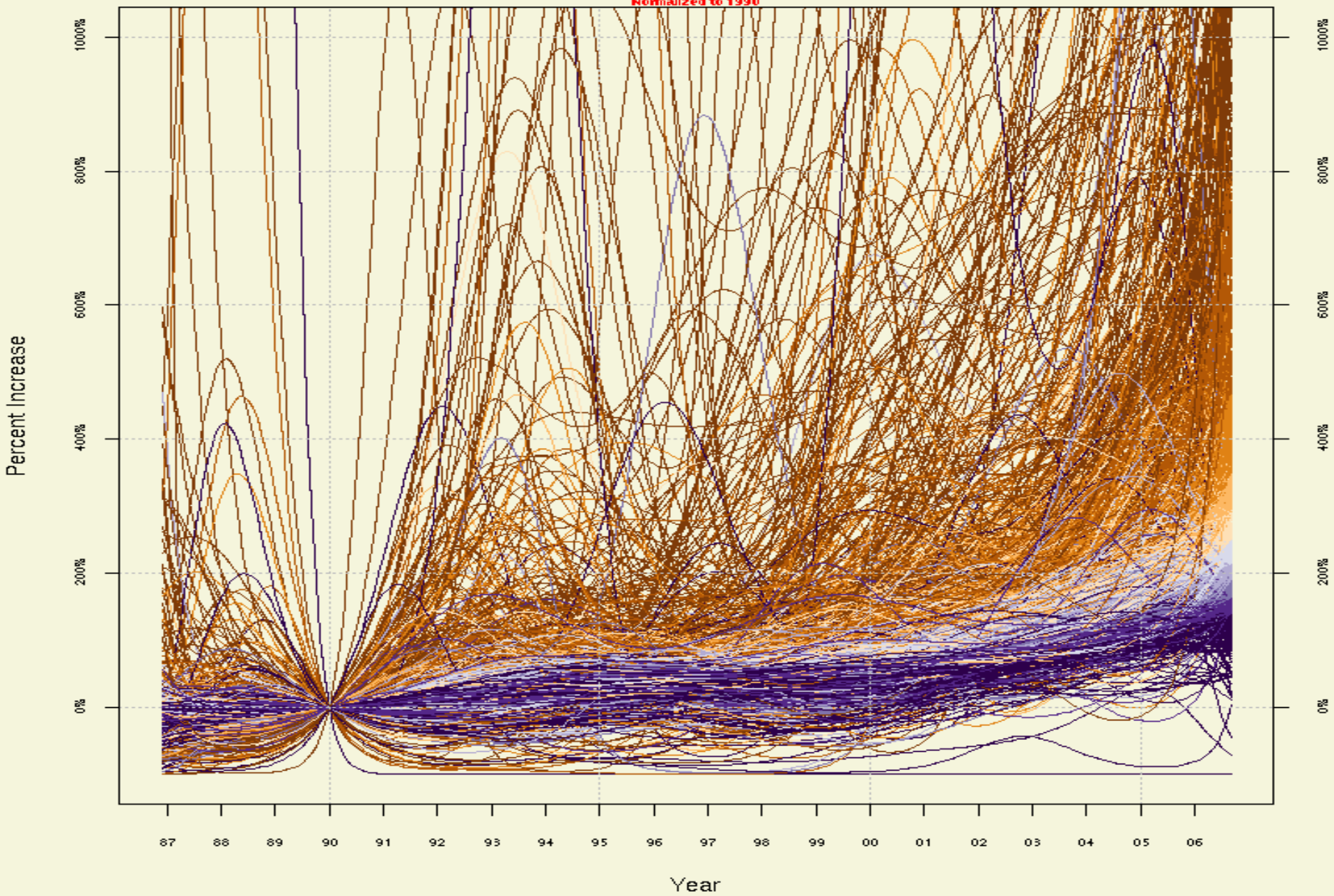
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Developing Tools: from Diagnostics to Investment



Chicago Tract Indices, Neighbor

Normalized to 1990



Identifying Patterns of Change

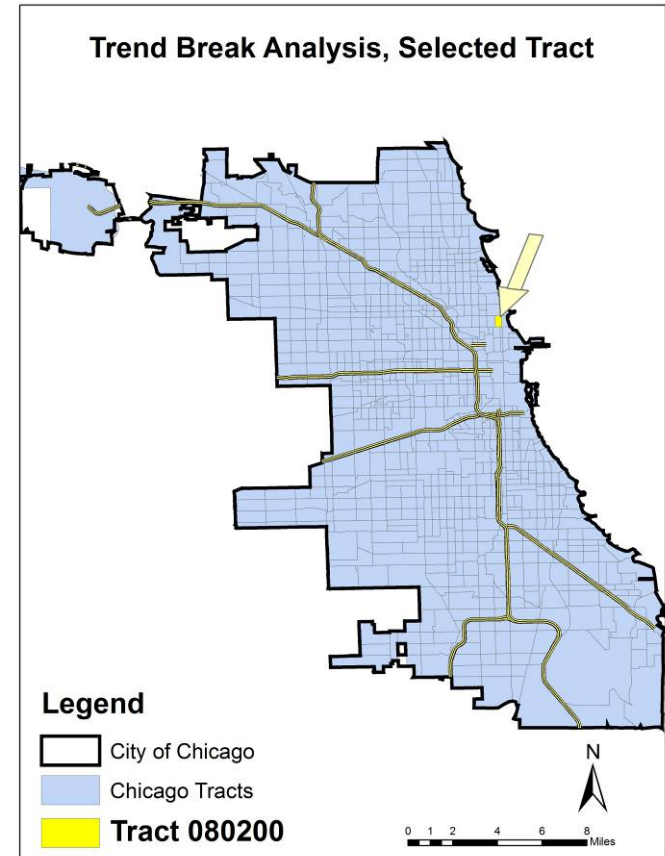
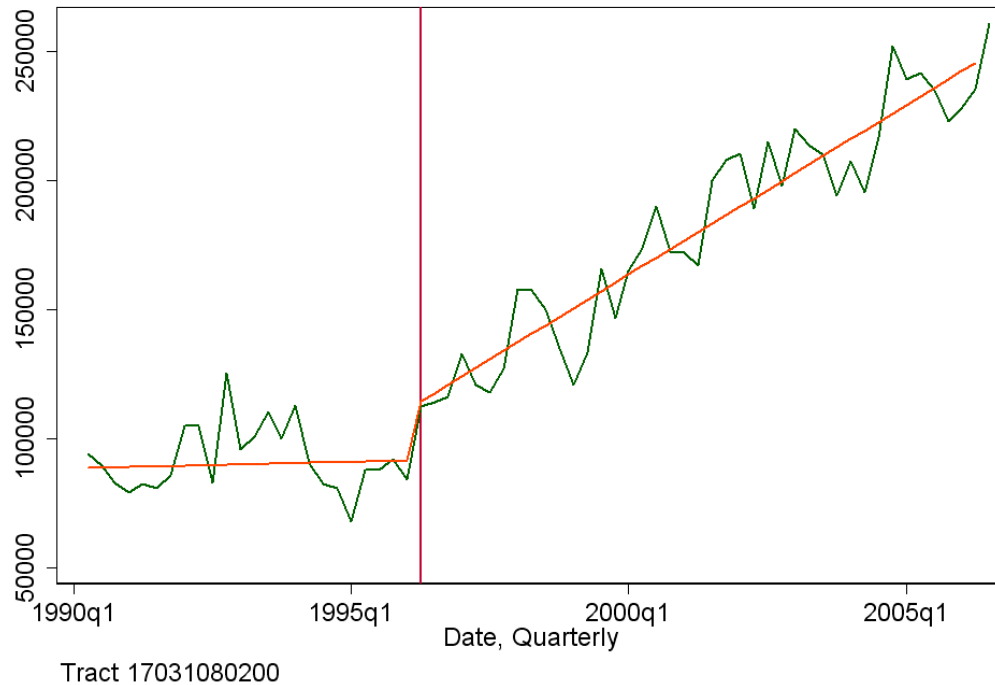
Three Complementary Methodologies:

- **Cluster Analysis**: group all neighborhoods by overall pattern
- **Trend Breaks**: classify neighborhoods based on number and type of structural breaks
- **Pattern Search**: specify a pattern of interest and search for matches in the data



Patterns of Interest: Tipping?

Chicago, North Side



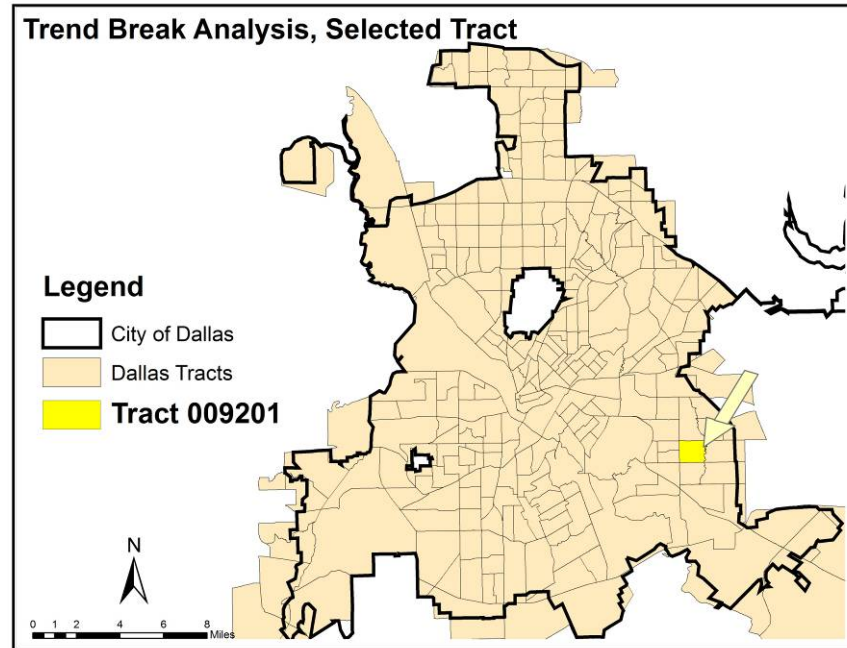
Statistically Identifying Structural Breaks

Patterns of Interest: Neighborhood Turnaround

Dallas, Southeast Side

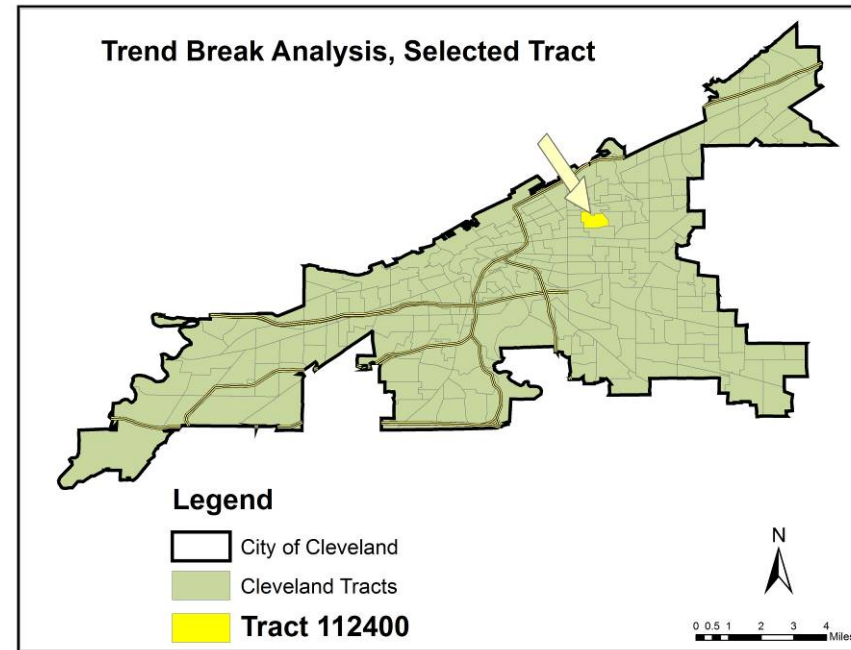
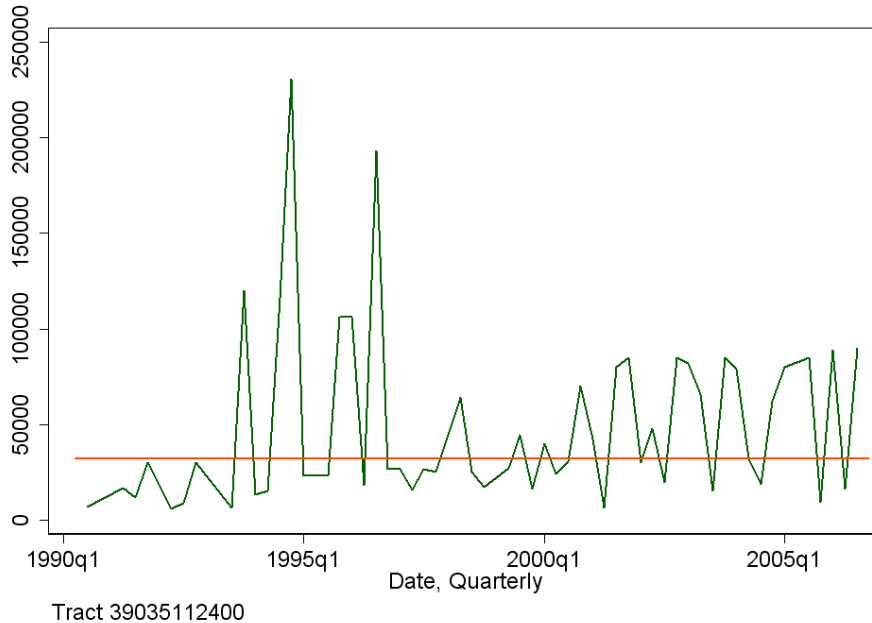


Tract 48113009201



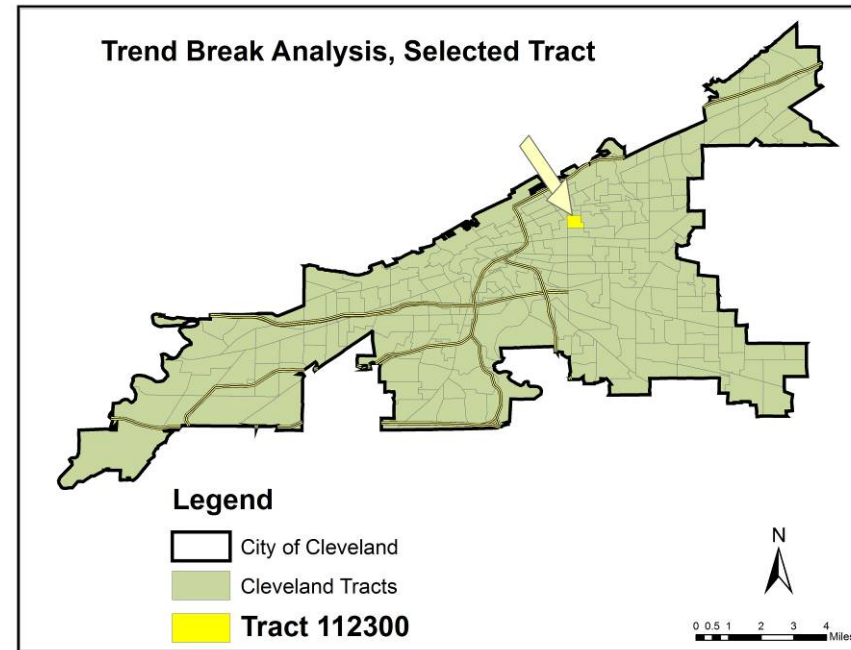
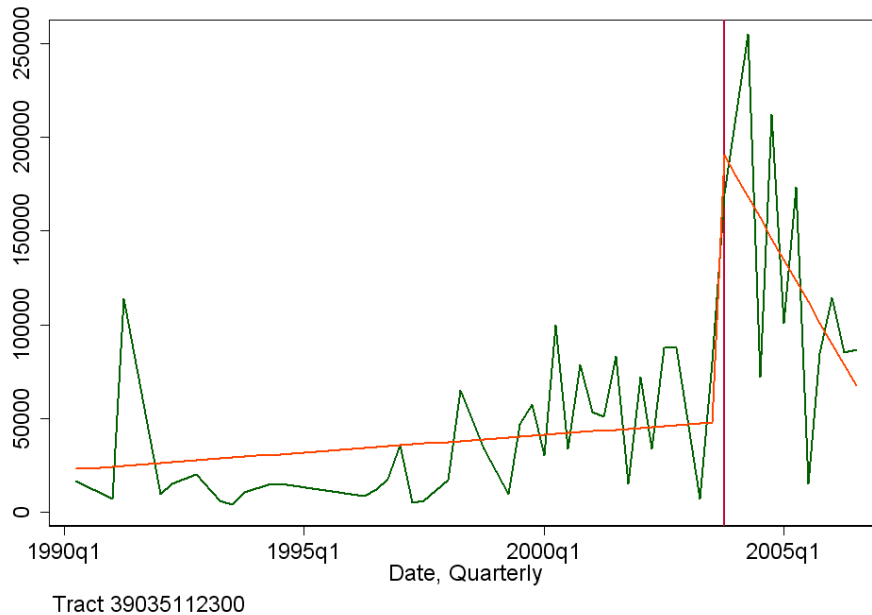
Patterns of Interest: Neighborhood Decline

Cleveland, East Side



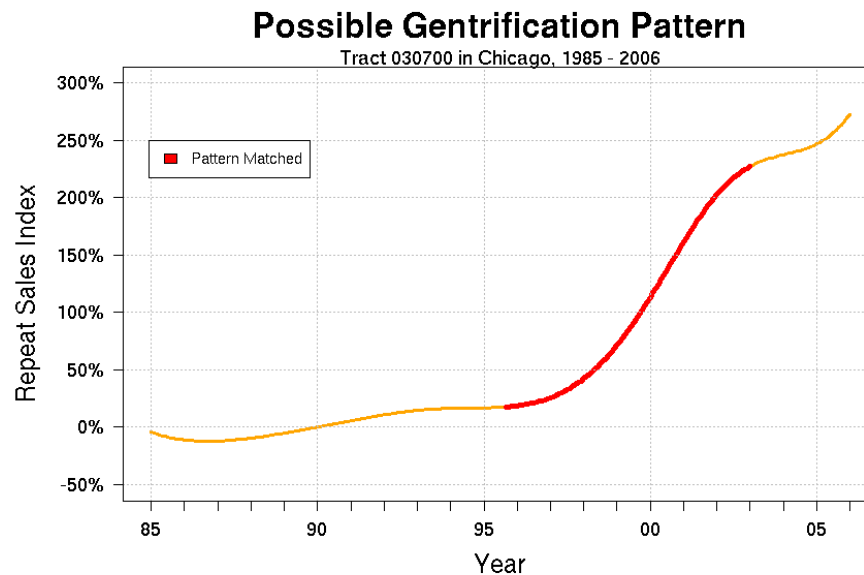
Patterns of Interest: Speculation?

Cleveland, East Side



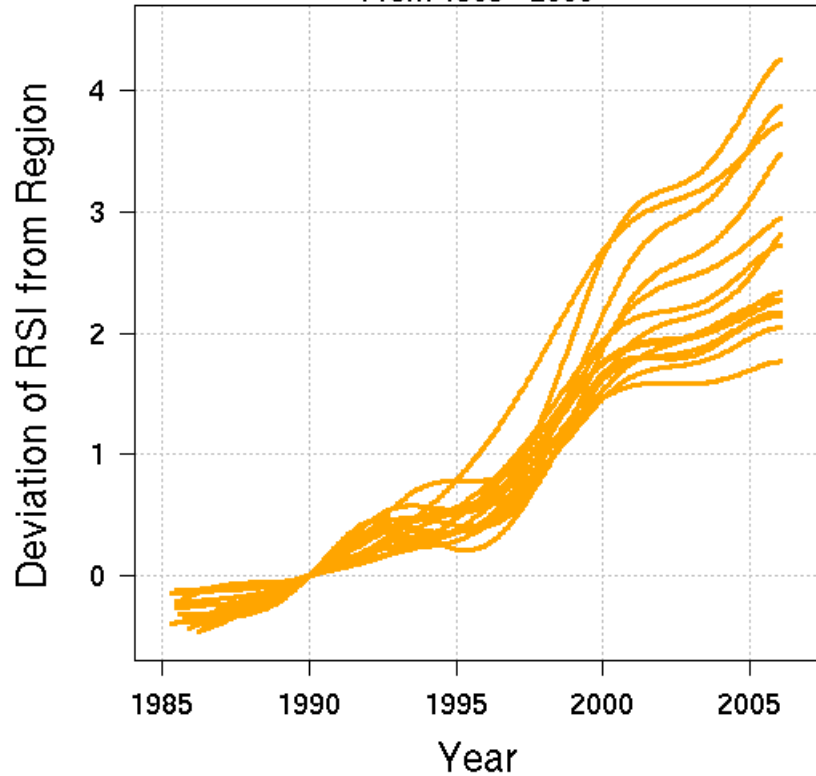
Pattern Search Example: Gentrification in Chicago

- **Goal:** Anticipating Neighborhood Change
- **How it Works:** Define a Pattern and Find Matching Cases
- **Example:** Possible Gentrification Pattern Defined Based on a Neighborhood in Chicago



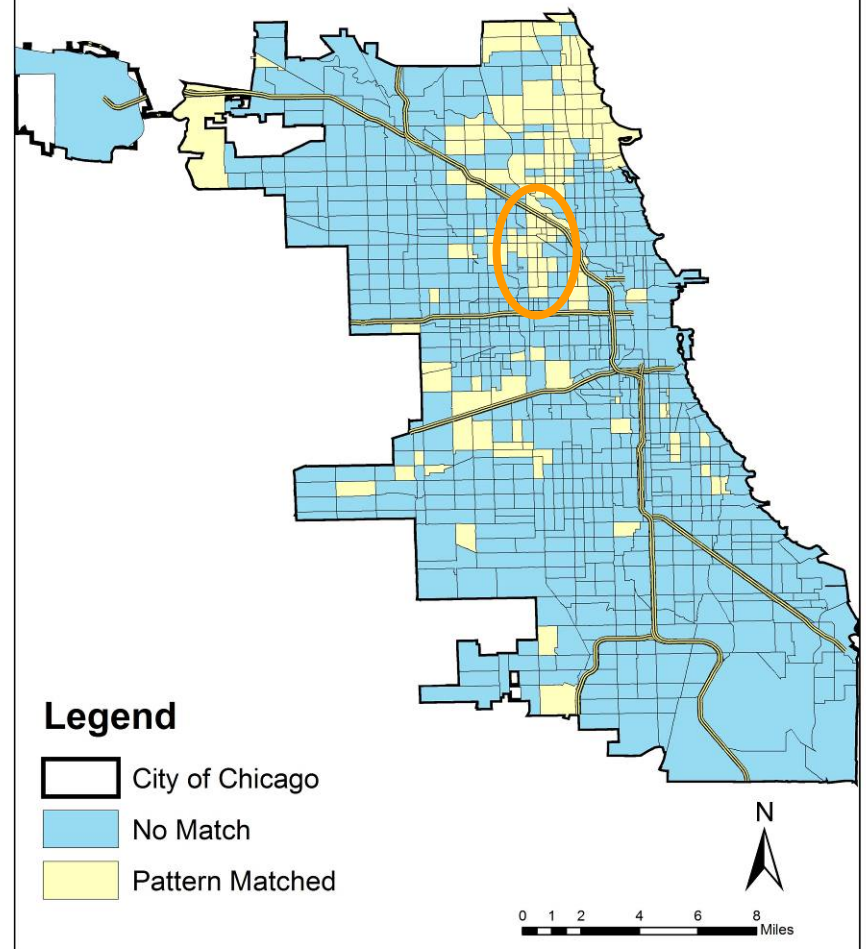
Zooming In: Wicker Park Area

All Tracts in Wicker Park
From 1985 - 2006

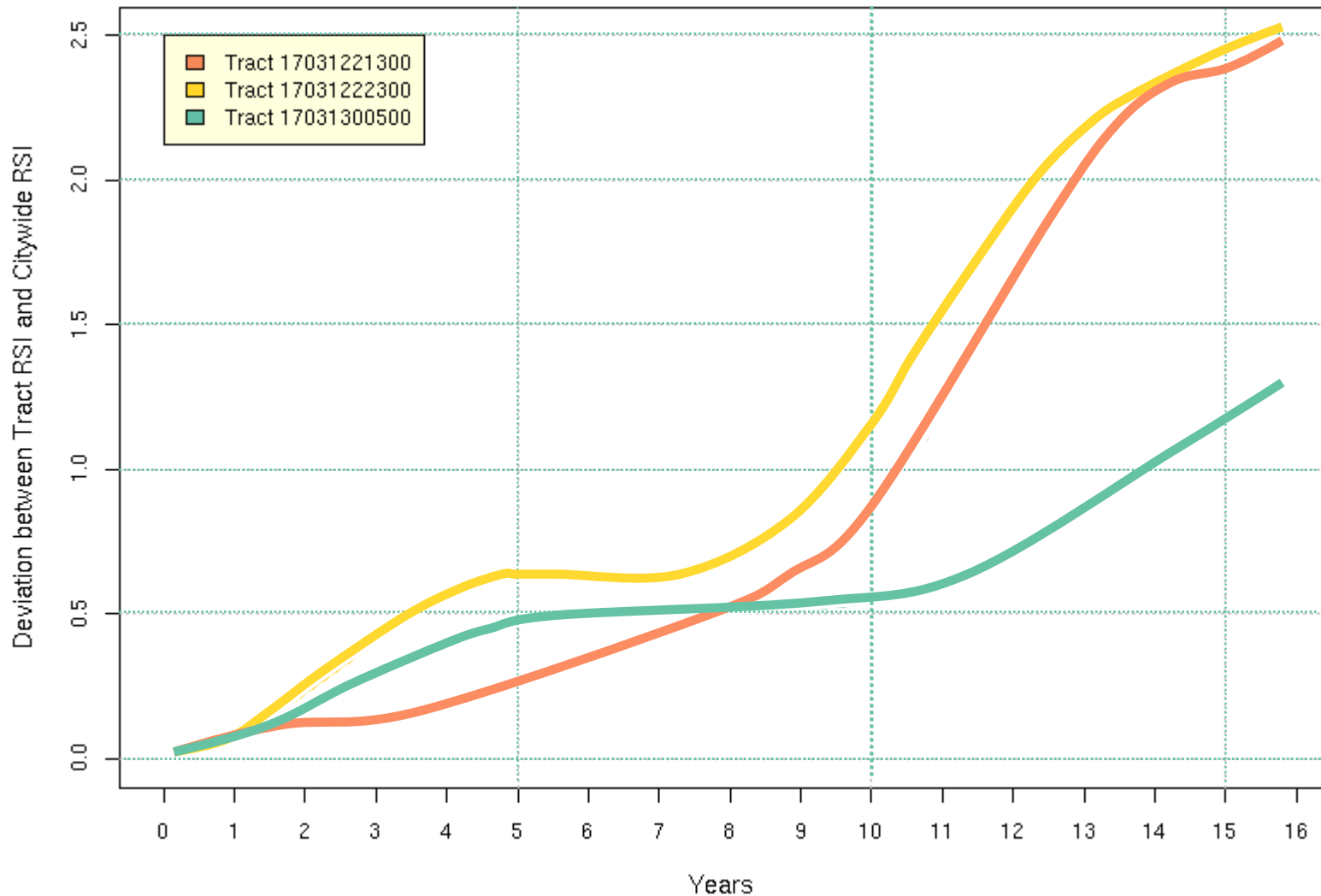


Possible Gentrification Pattern

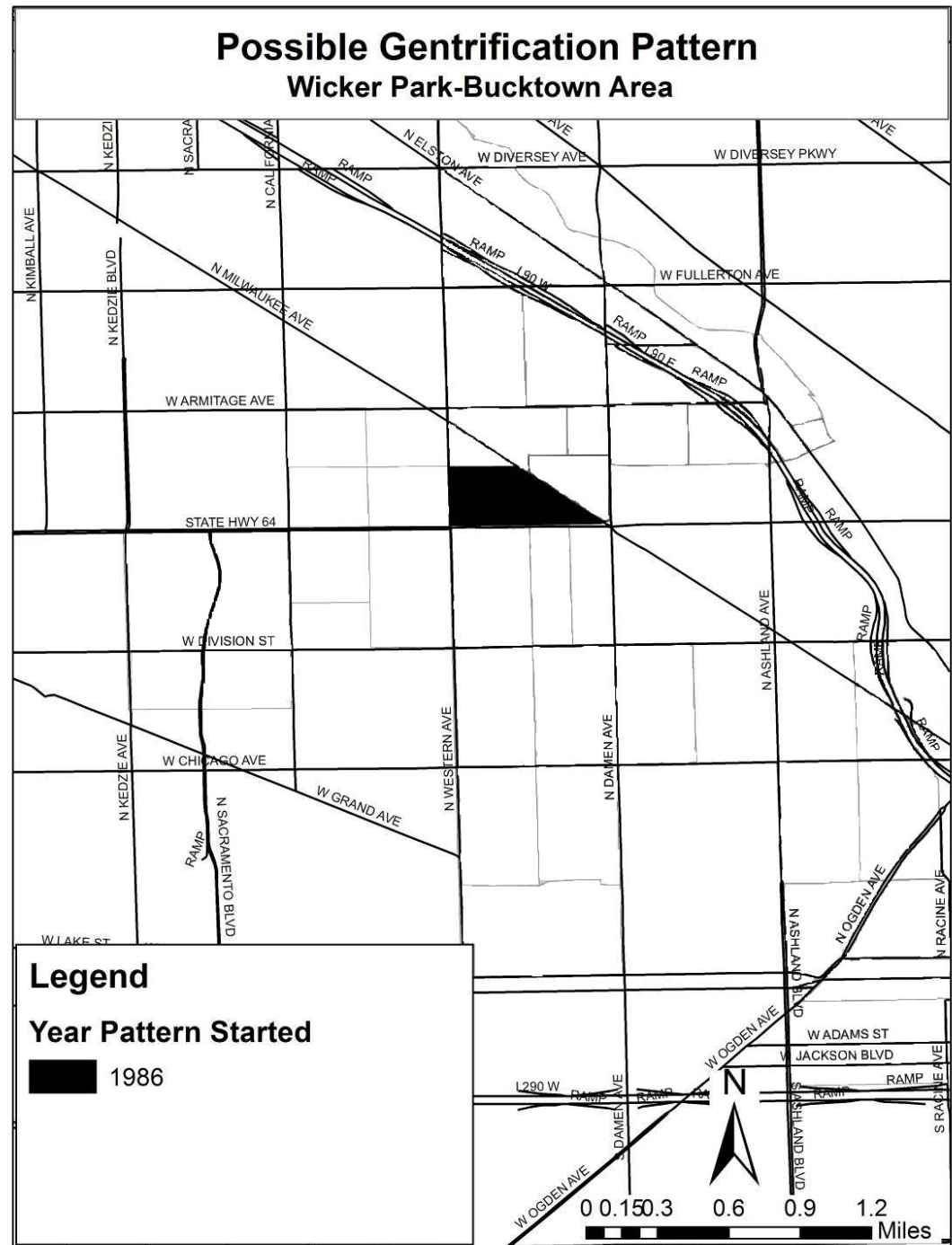
Chicago Tracts, 1985-2006



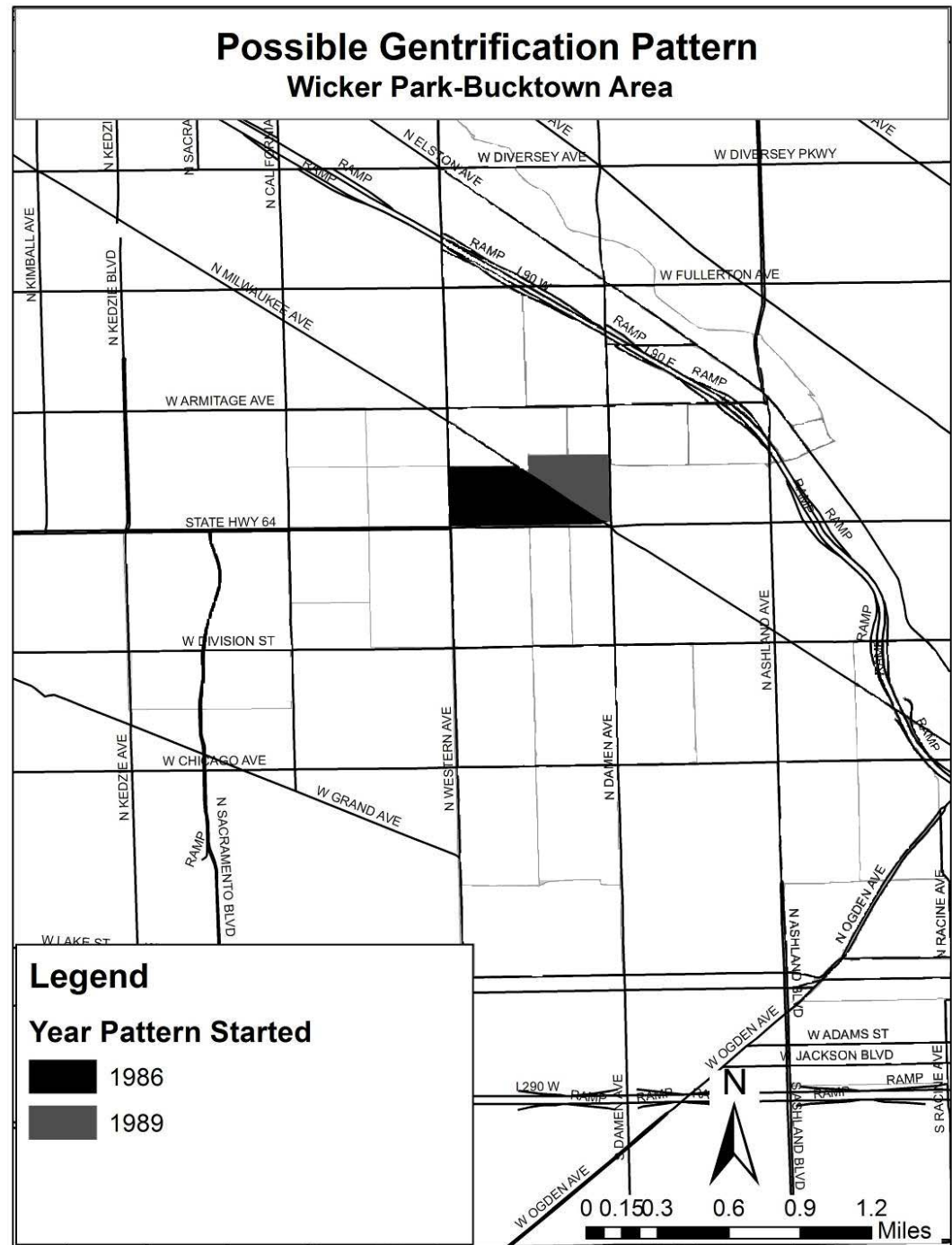
Possible Application: Anticipating and Managing Gentrification



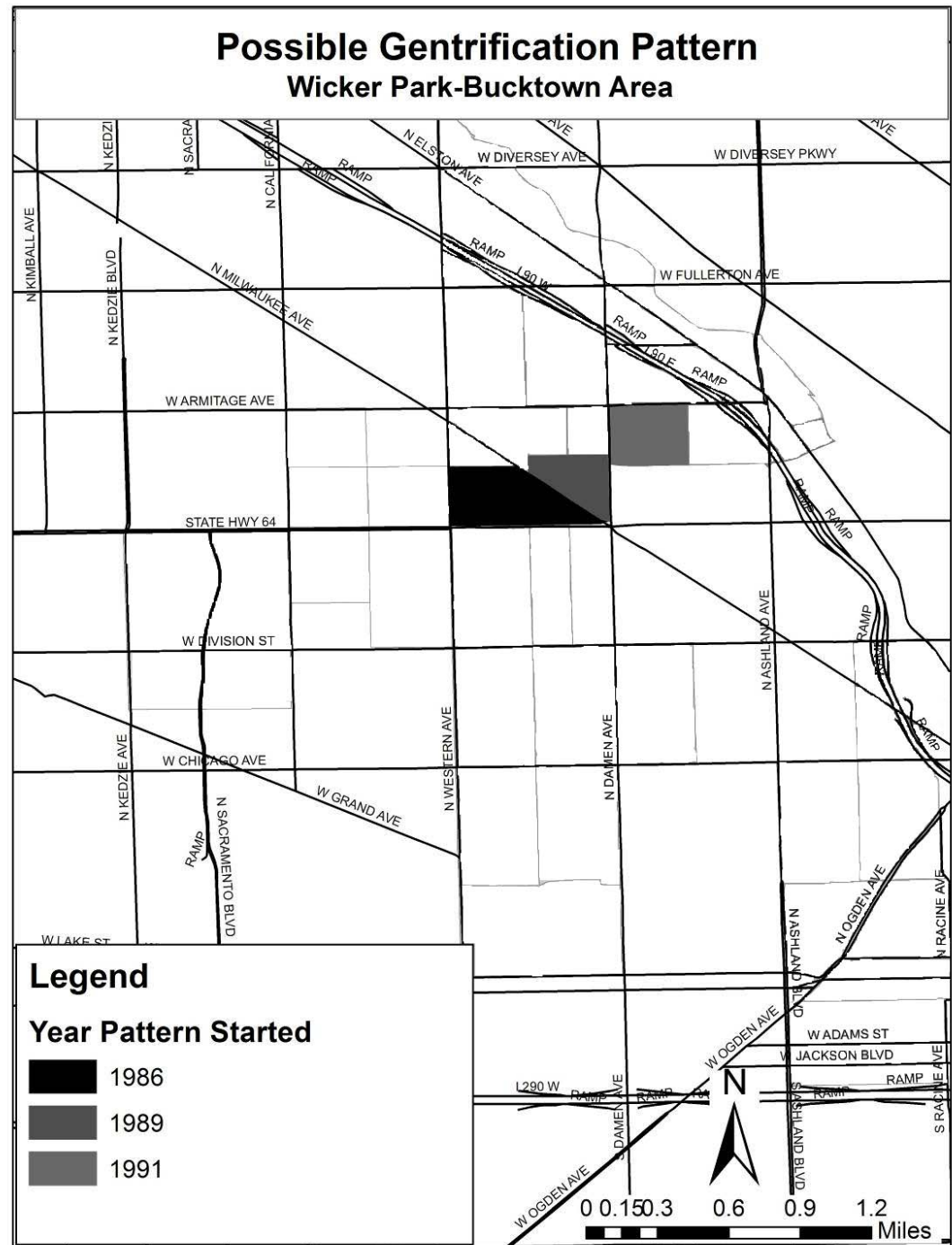
Pattern “Spreading” to Nearby Tracts



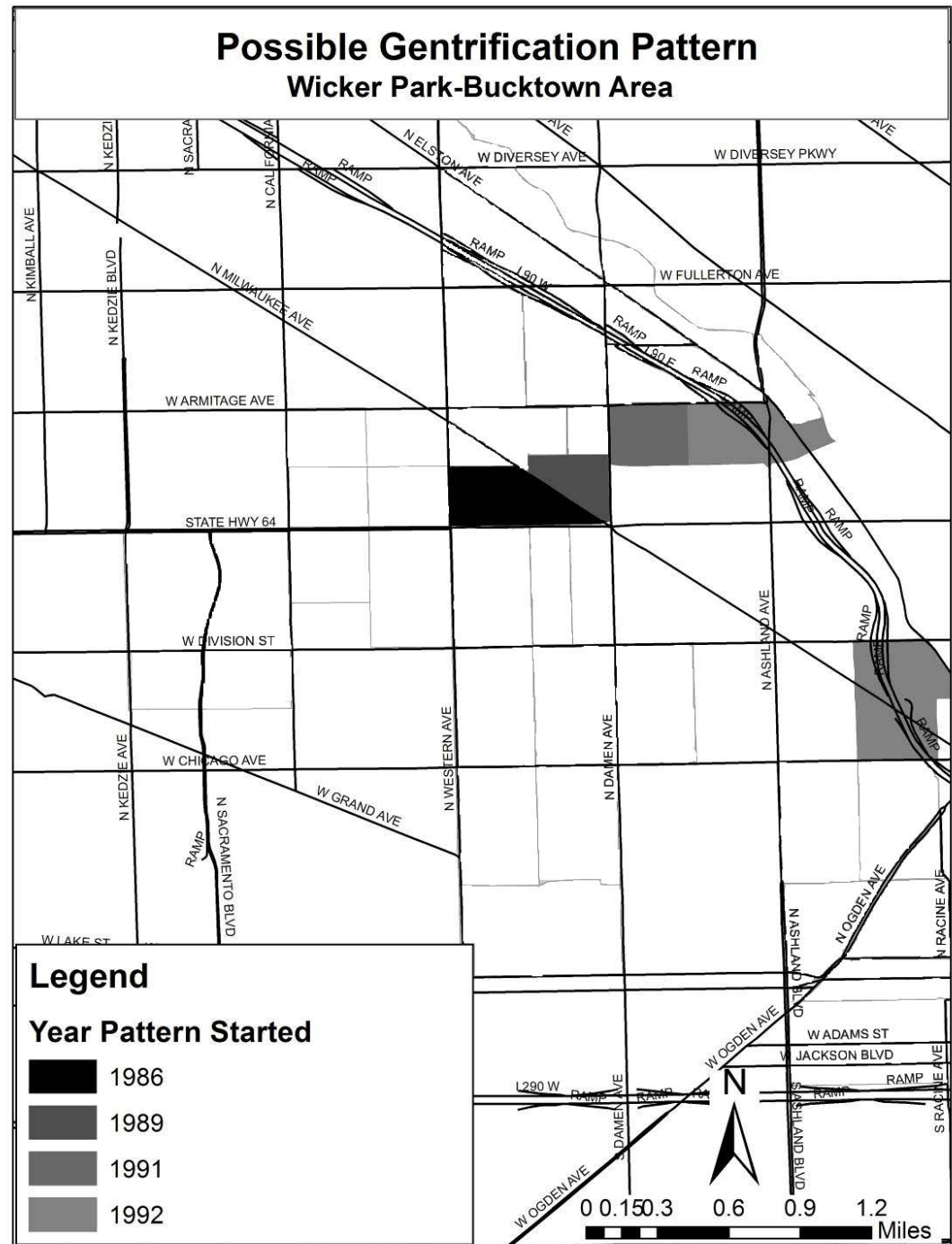
Pattern “Spreading” to Nearby Tracts



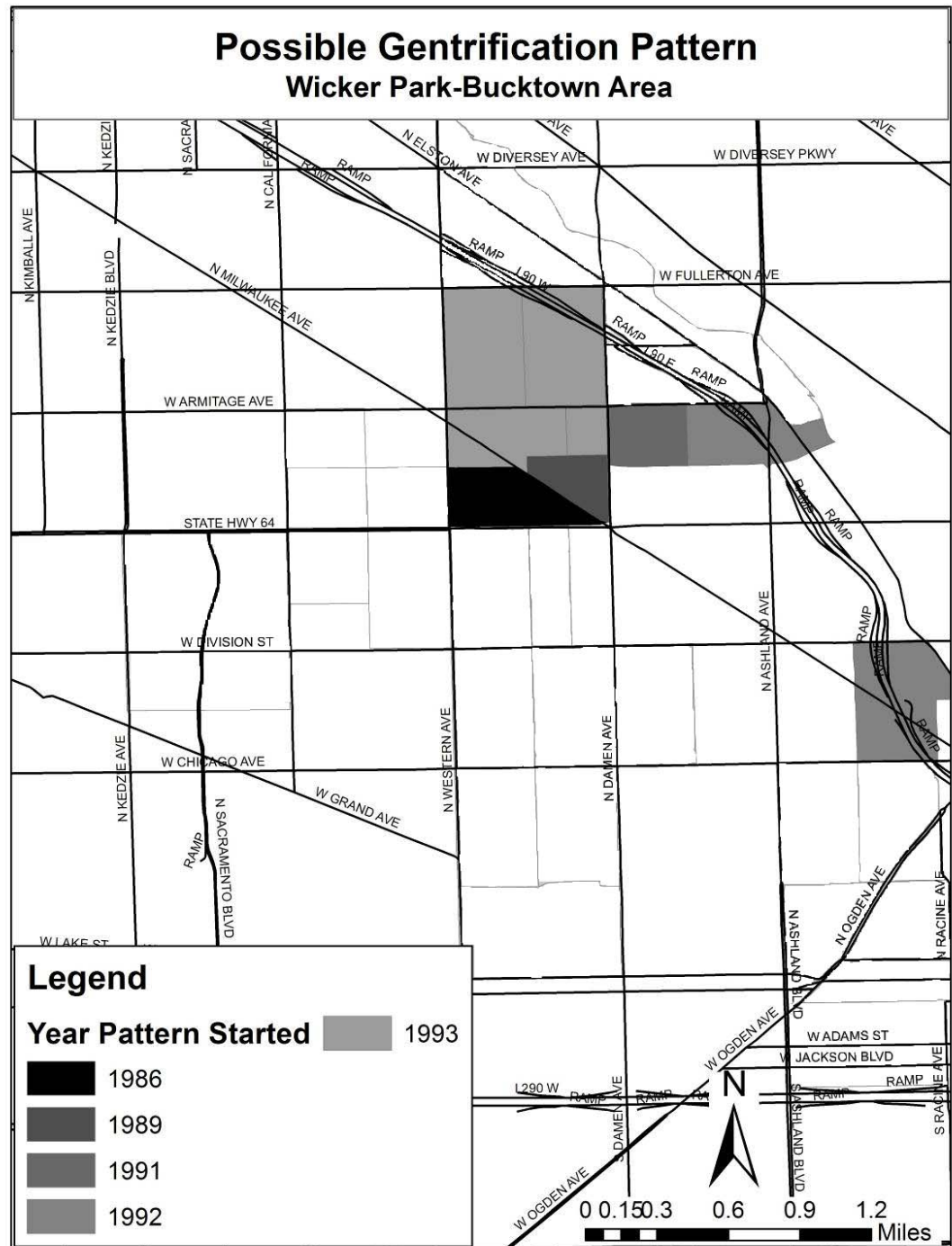
Pattern “Spreading” to Nearby Tracts



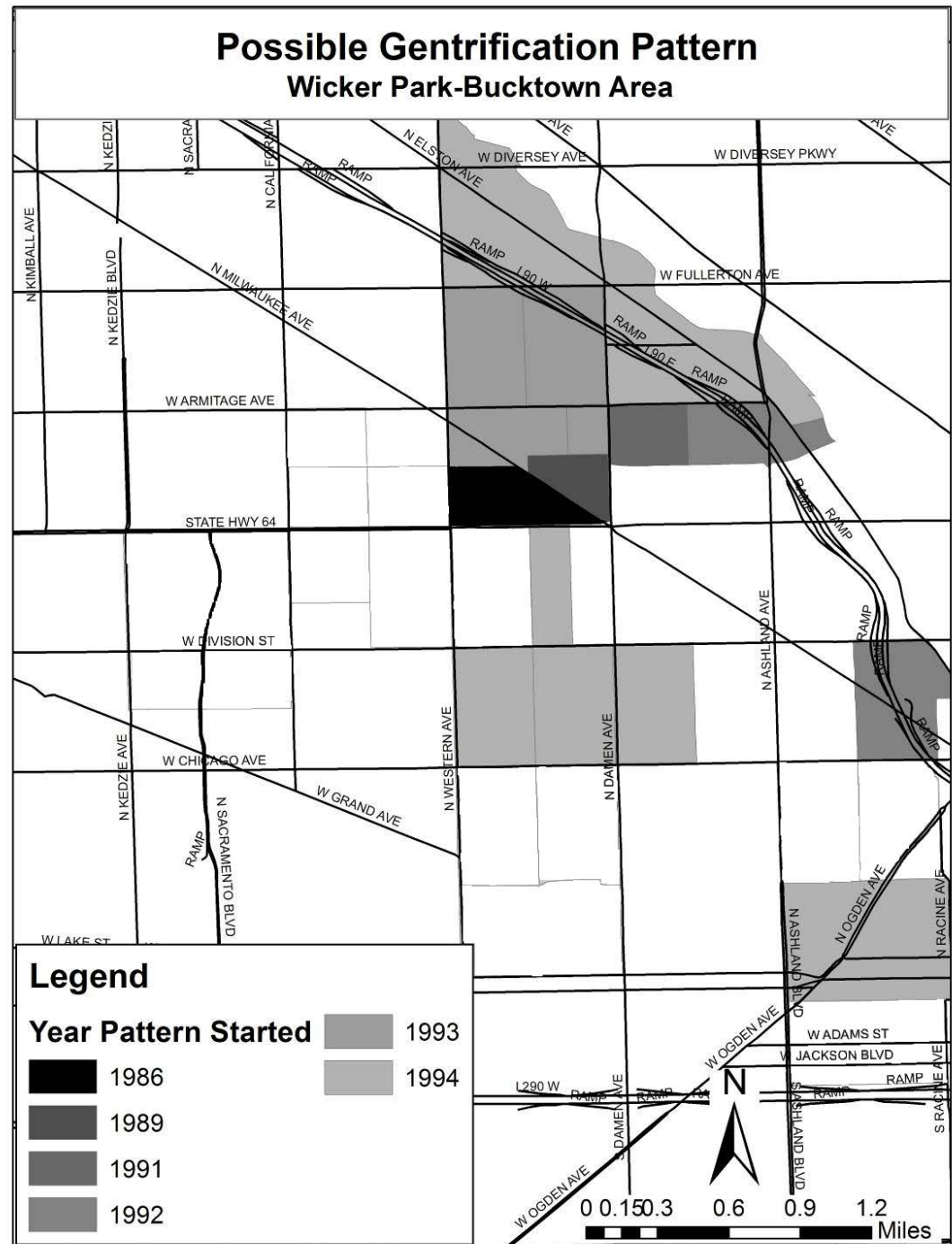
Pattern “Spreading” to Nearby Tracts



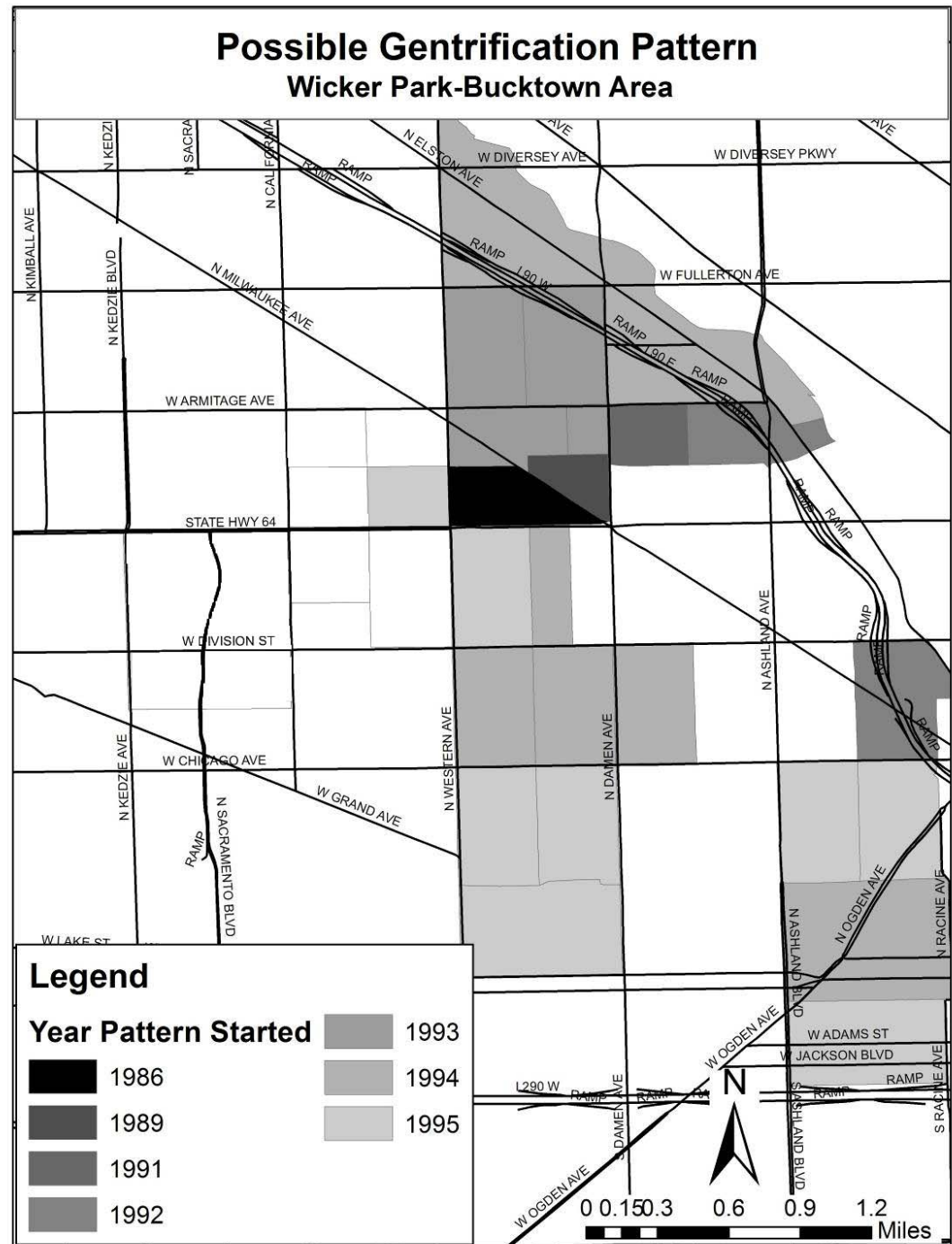
Pattern “Spreading” to Nearby Tracts



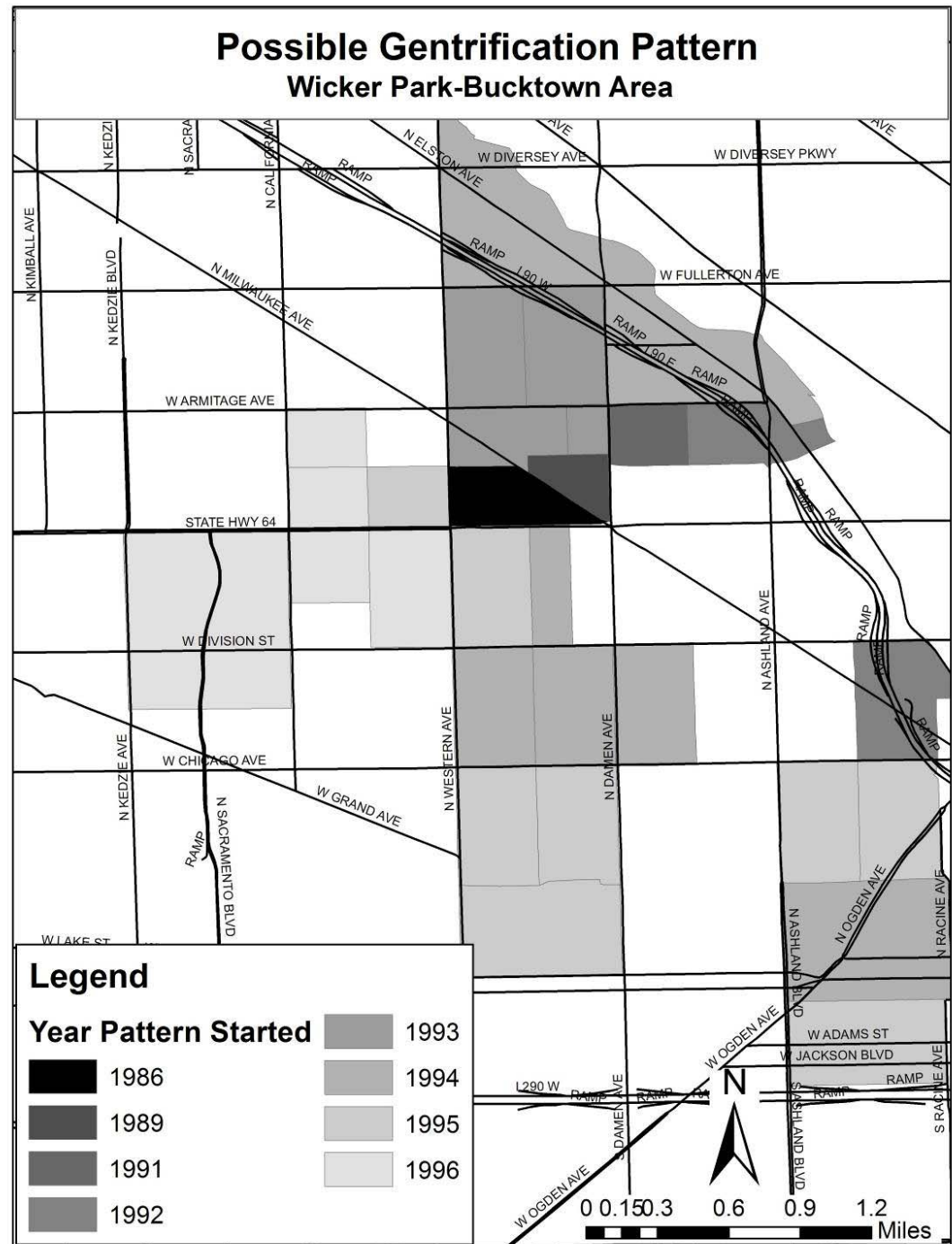
Pattern “Spreading” to Nearby Tracts



Pattern “Spreading” to Nearby Tracts



Pattern “Spreading” to Nearby Tracts



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Developing Tools: from Diagnostics to Investment



Developing New Tools for the Field

Question/Goal	Tool
Enabling Investment in Inner City Real Estate Markets	RSI → REIT
Track Affordability and Neighborhood Housing Mix	Housing Diversity Metric
Anticipate and Manage Gentrification	Early Warning System
Planning Community Development Interventions	Neighborhood Change Simulation
What neighborhoods are similar along multiple dimensions of interest?	Similarity Index/ Custom Typology
What drivers differentiate neighborhoods with respect to a specific outcome of interest?	CART
How will a specific intervention affect its surrounding area?	Impact Estimator
What locations will maximize the impact of an intervention?	Spatial Multiplier
What is my “real” neighborhood?	Semivariogram

Housing Diversity Metric

What It Does:

- Tracks the affordability and mix of the housing stock (distribution, not just median)

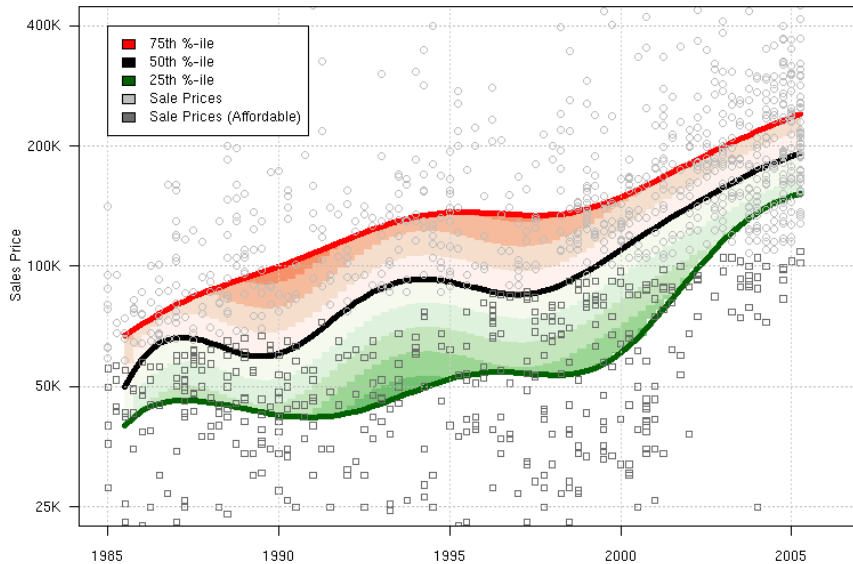
Applications:

- Enables tracking the range of housing available in the neighborhood
- Better indicator of possible displacement than median prices alone



Example: Tracking the Price Mix

Sale Prices at 25th, 50th, and 75th Percentiles of tract 17031010200 in Chicago

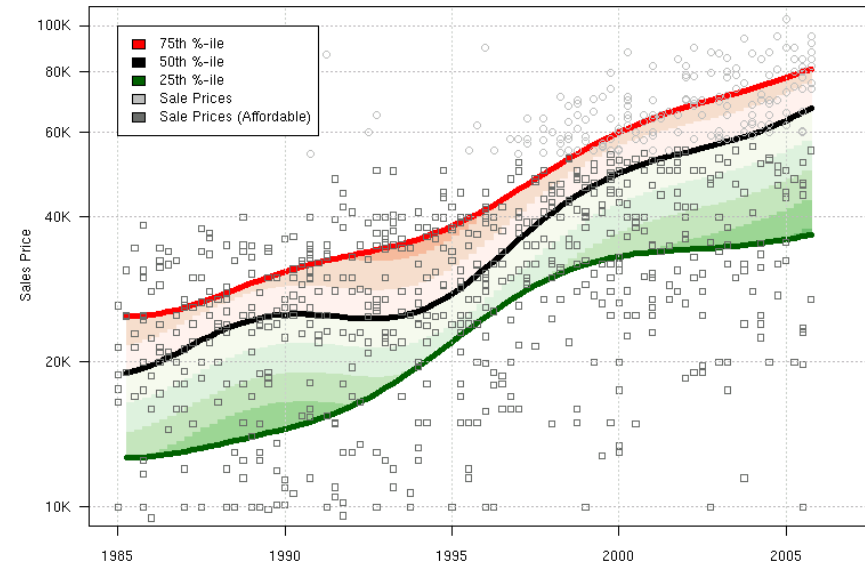


**Strong Overall Appreciation,
Range of Housing Options Is Narrowing**



Lack of Affordable Housing

Sale Prices at 25th, 50th, and 75th Percentiles of tract 39035105100 in Cleveland



**Strong Overall Appreciation, but
Range of Housing Options Is Still Wide**



**Large Share of Housing
Remains Affordable**

Classification and Regression Tree (CART)

What It Does:

- Identify similar neighborhoods with respect to an outcome of interest and its drivers

Applications:

- Identify leverage points to affect the desired outcome
- Meaningful comparison of trends and best practices across neighborhoods

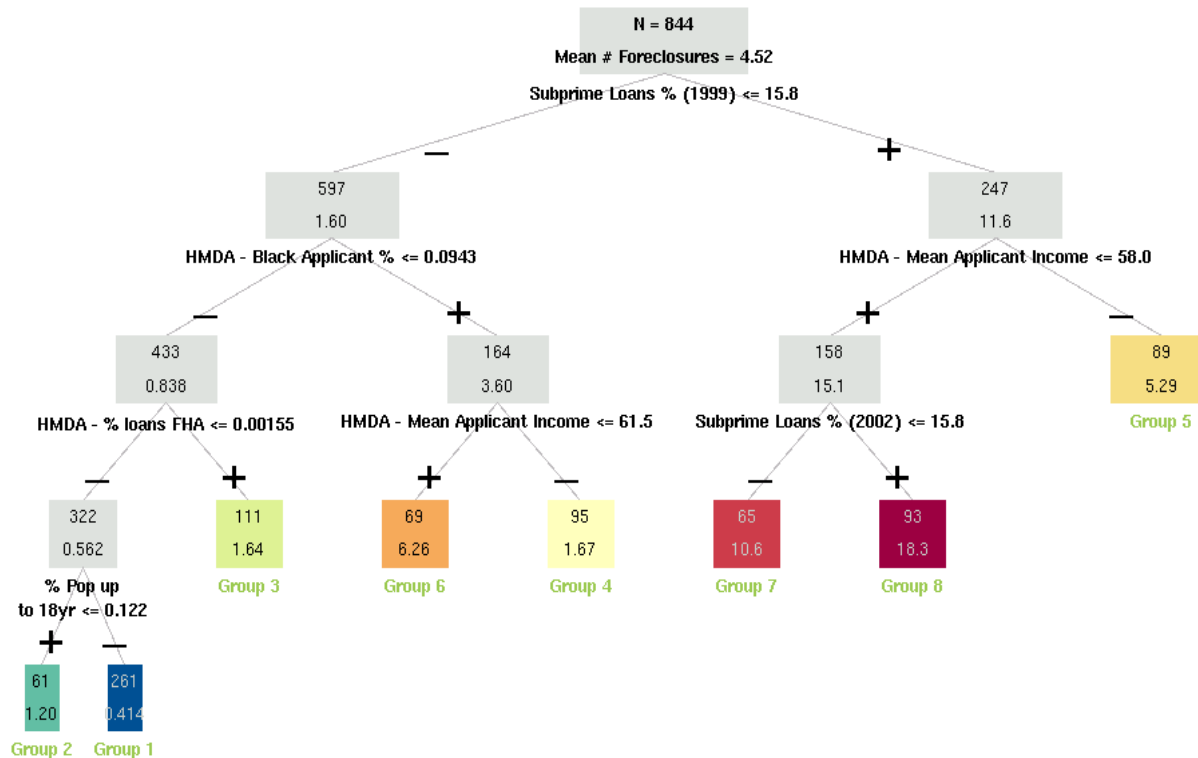


Sample CART: Foreclosures

CART Tree built using "Number of Foreclosures in 2004" as the Dependent Variable

All Variables Measured in 2004 unless otherwise noted

40 Variables Tested



Overall Fit: R2 = 0.378

+/- signs indicate which side has greater/less mean value

Groups are color coded by # foreclosures in a color spectrum: highest = Red -> Yellow -> Green -> Blue = lowest

Outcome:

- Number of Foreclosures (2004)

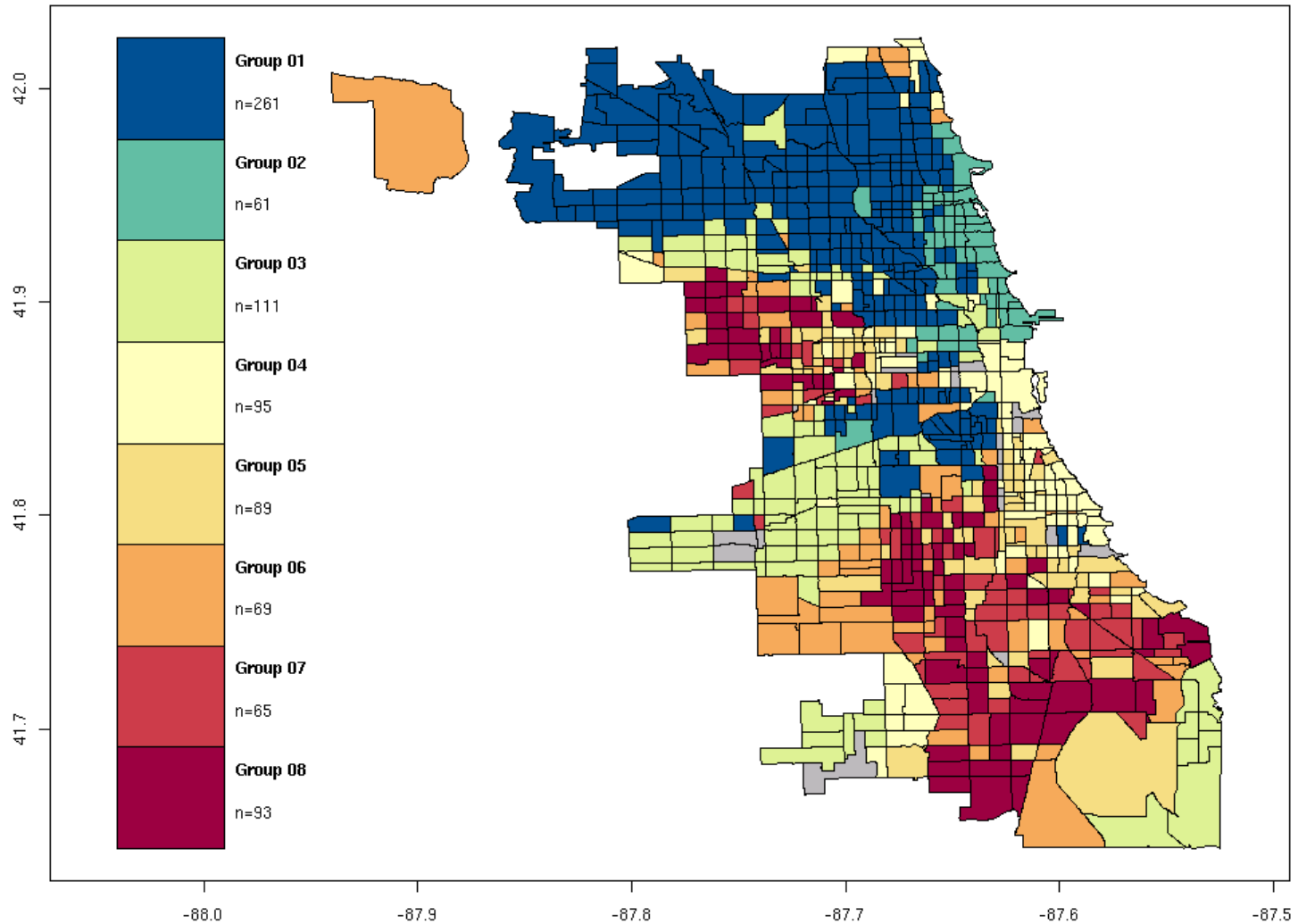
Drivers:

- % Subprime Loans in Previous Years
- Mean Loan Applicant Income
- % FHA Loans
- % Black Borrowers

What Neighborhoods Have Similar Numbers of Foreclosures, and Why?

CART Output: Chicago Segments

Geographic Distribution of CART Groups in Chicago, using Foreclosures as the Dependent Variable



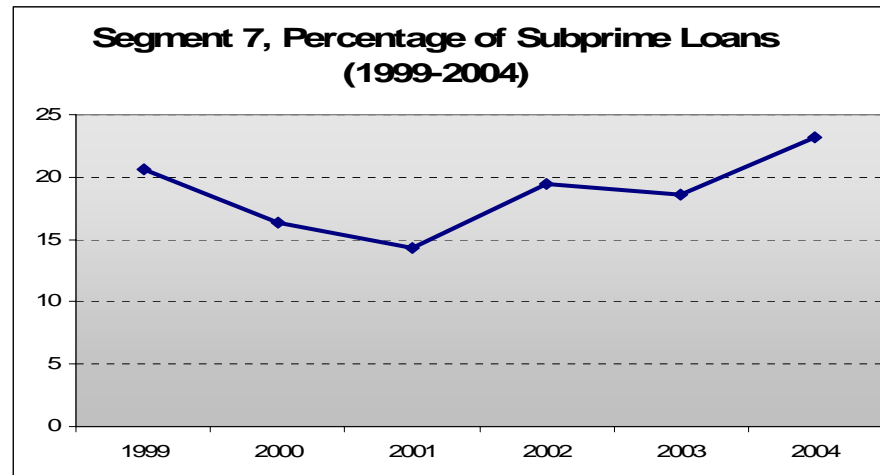
Cluster 7: Defining Traits and Risk Factors

Segment Profile:

- Isolated, underserved, predominantly African American communities. High rates of unemployment and subprime lending activity.

Primary Risk Factor:

- Percentage of subprime loans (primary driver of foreclosures) is at its highest and still on the rise



Impact Estimator

What It Does:

- Estimate impact of an intervention on surrounding housing values (or on other outcome, e.g. crime)

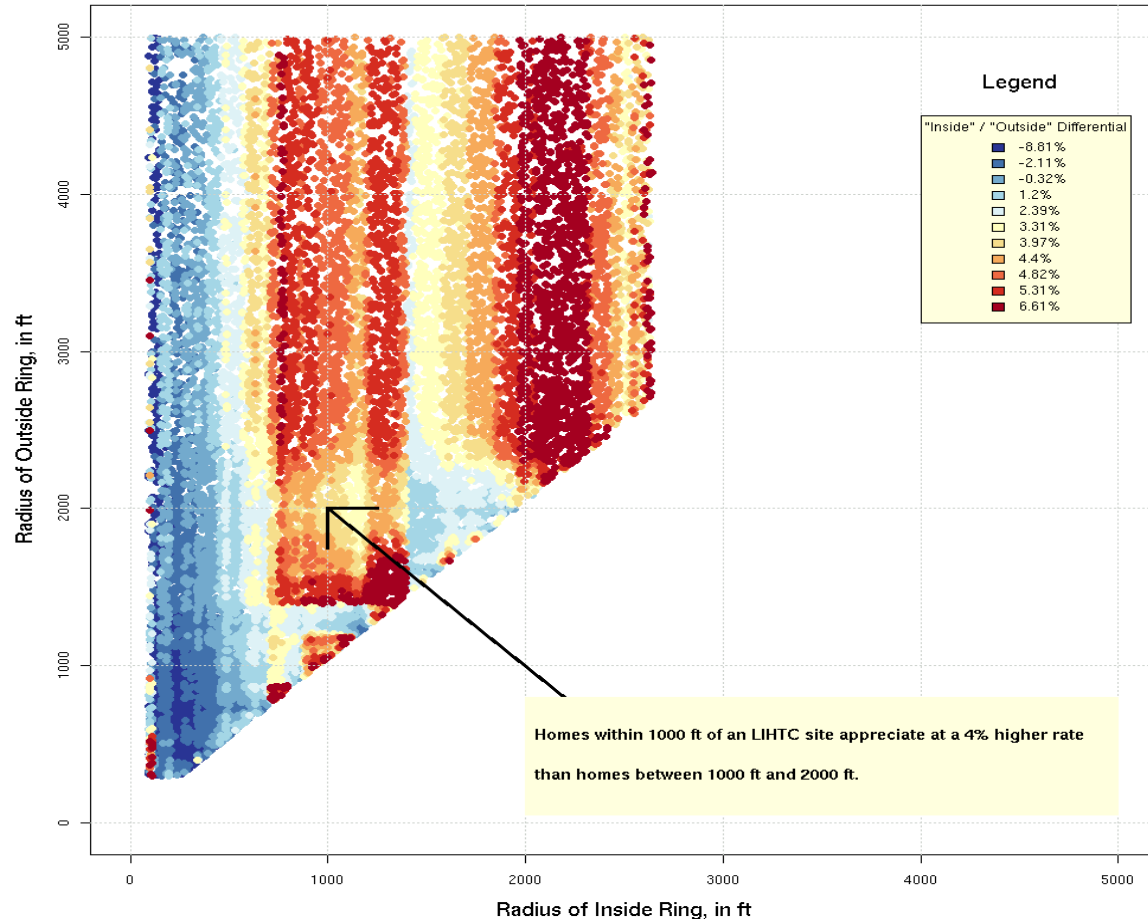
Possible Applications:

- Evaluate the impact of a development policy
- Choose among alternative interventions based on estimated benefits to the surrounding community
- Advocate for a specific intervention



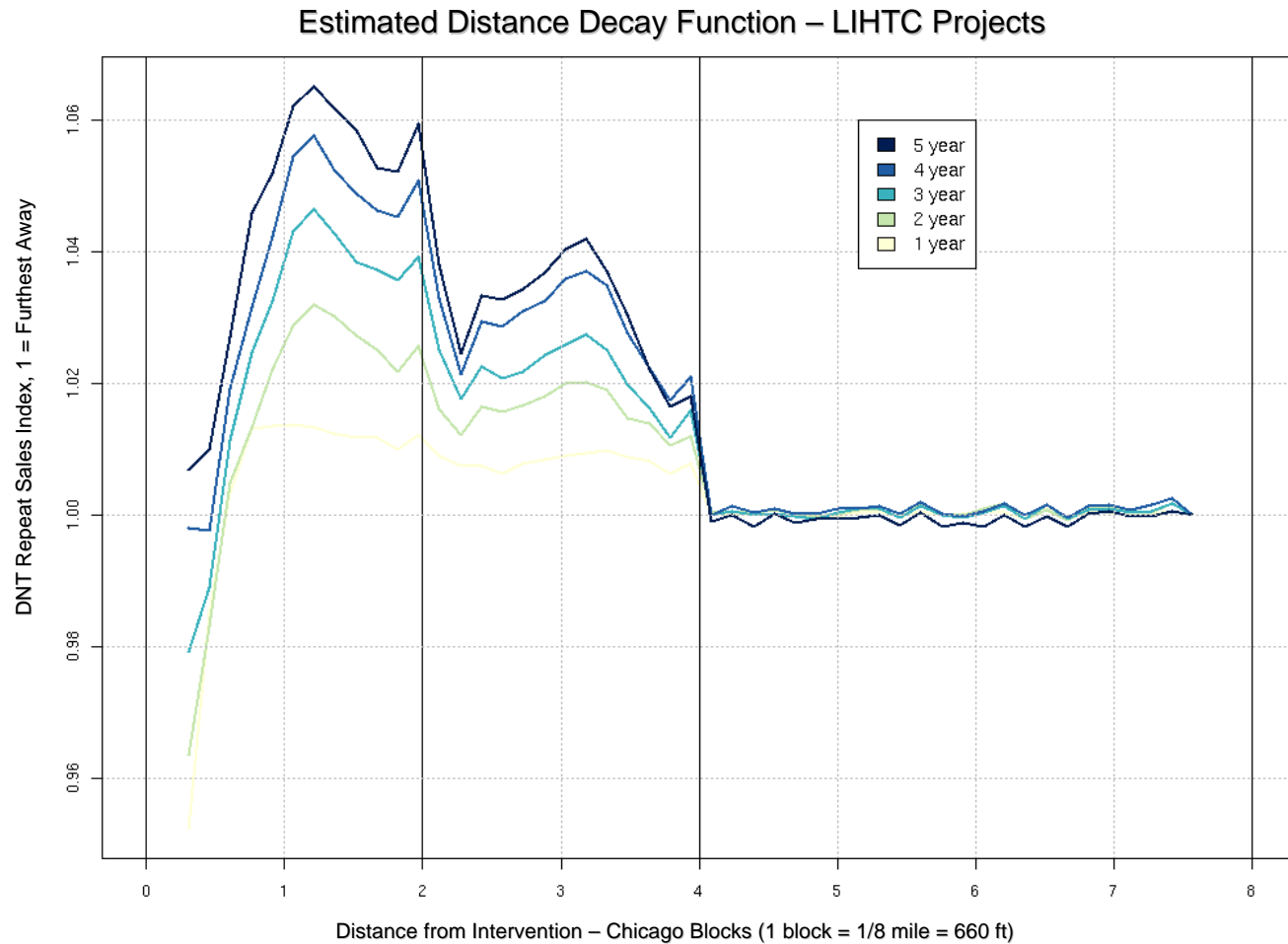
Example: What is the effect over time and space of LIHTC housing?

Comparing the Distance Effects of LIHTC projects on Local Housing Appreciation

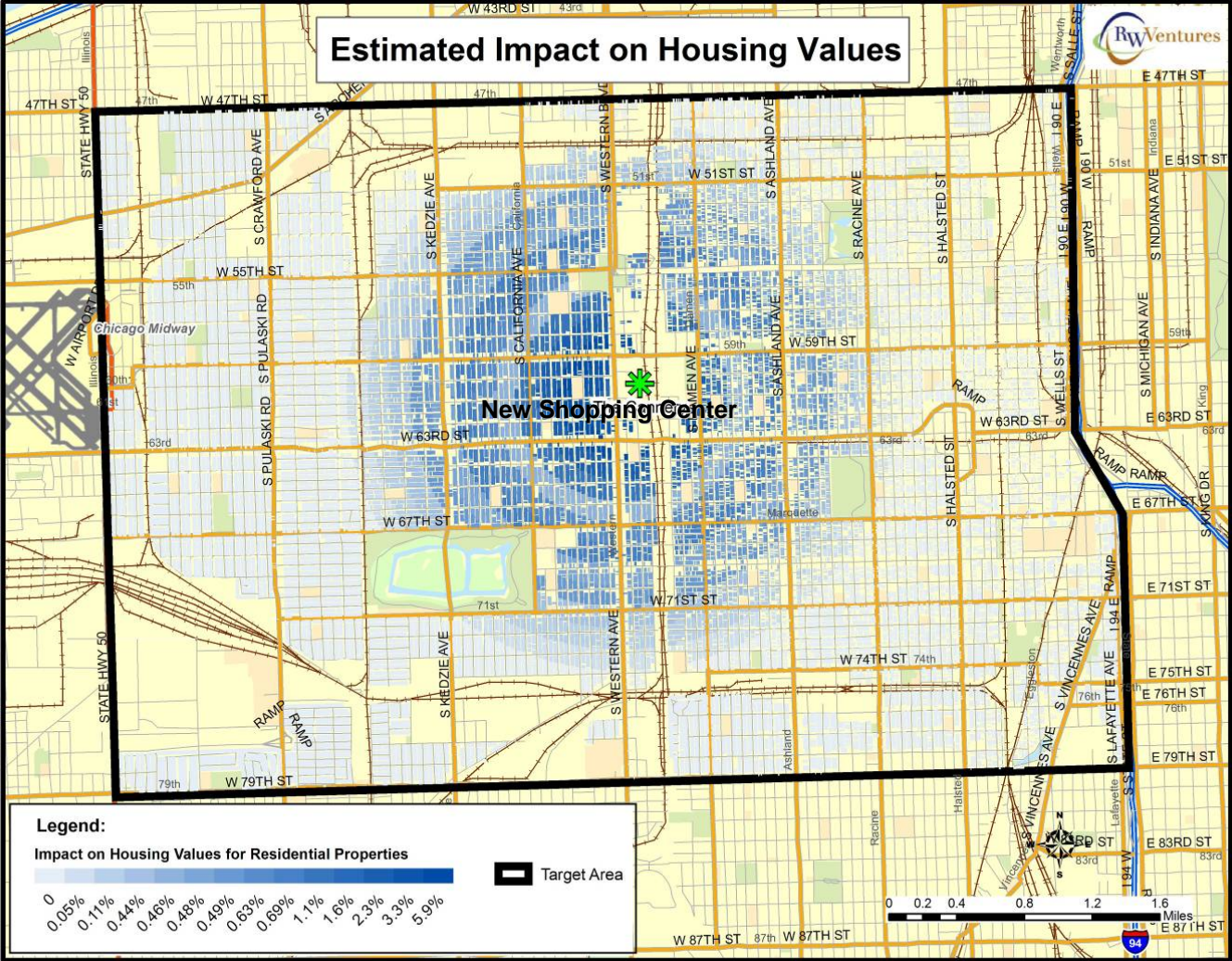


Monte Carlo Simulation to Estimate Impact Variation with Distance

Impact of LIHTC on Surrounding Properties



Applying the Estimator to a Specific Project: New Shopping Center in Chicago



Estimated benefits to the community: \$29 million in increased property values, or an average of \$1,300 per home owner

Ongoing and Inclusive Process

- **Positioning in the Field**
 - Project based on learning from other initiatives
 - Results intended to contribute to their work
- **Ongoing Process**
 - Project is iterative
 - Results need to be used and continually refined
- **Inclusiveness**
 - Multiple partners in various roles
 - Open Source



Discussion

- General Comments and Questions?
- Patterns of Change of Particular Interest?
- What are People Trying to Better Understand About Neighborhoods?
- What Tools and Applications Would Be Most Useful?
- Partners: Corollary Research, Tool Development and Testing, Other?



Dynamic Neighborhood Taxonomy

For more information, please visit:

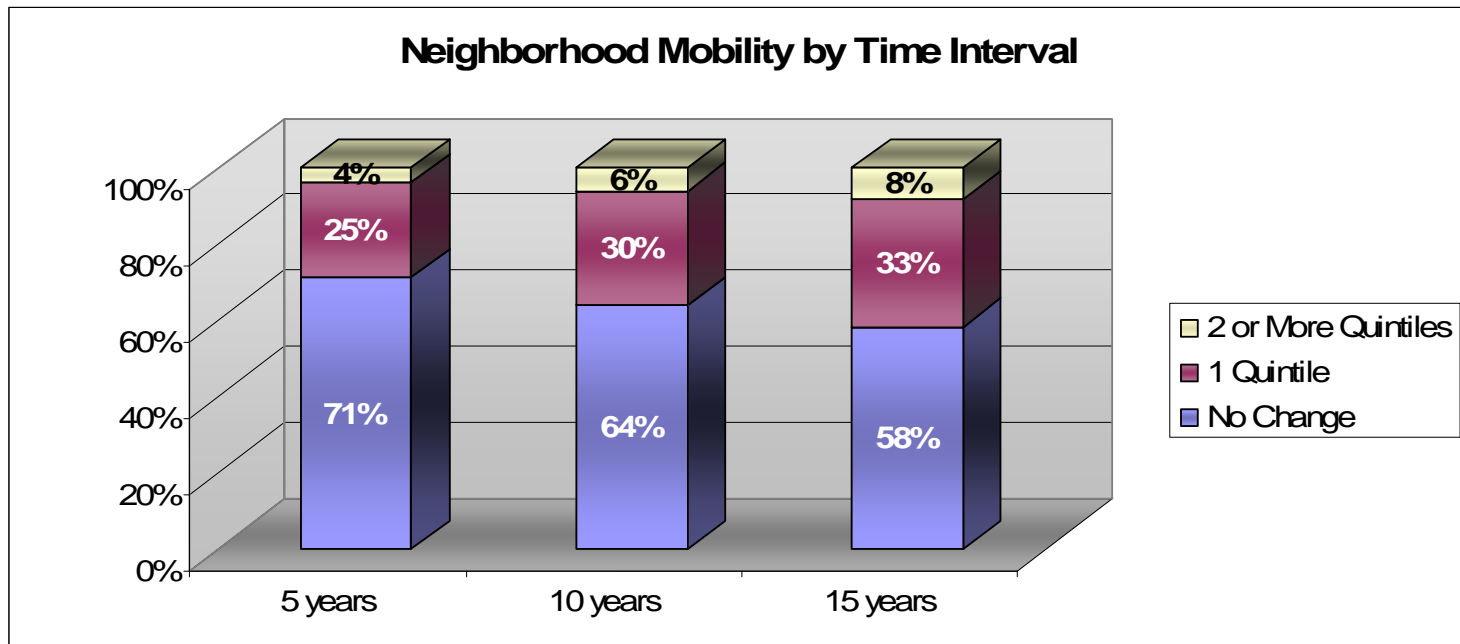
www.rw-ventures.com/RWteam

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by
RW Ventures, LLC



Neighborhood Change Is a Slow Process



Over 15 years, most neighborhoods do not change their position relative to other neighborhoods in the region.

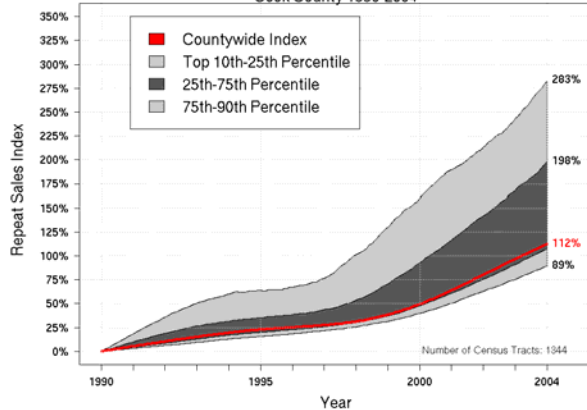
Target Analysis to Neighborhoods with Different Degrees of Change

Median Sales Price Transition Matrix Cleveland, 1990-2004					
	Final Quintile				
Initial Quintile	1	2	3	4	5
1	76.9%	15.4%	7.7%	0.0%	0.0%
2	5.1%	51.3%	25.6%	15.4%	2.6%
3	2.6%	26.3%	26.3%	39.5%	5.3%
4	7.7%	2.6%	28.2%	23.1%	38.5%
5	7.7%	5.1%	10.3%	23.1%	53.8%

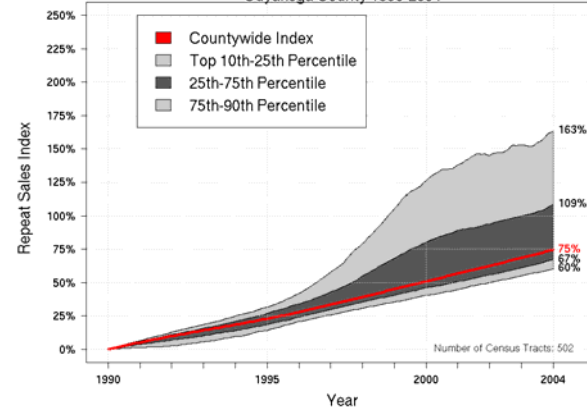
In Cleveland, 13% of all the tracts at the bottom of the distribution in 1990 moved up to the top 2 quintiles 15 years later.

Neighborhoods and Regions

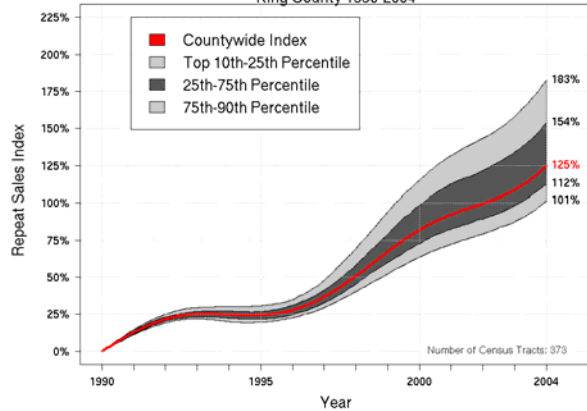
Variation of appreciation trends across Neighborhoods
Cook County 1990-2004



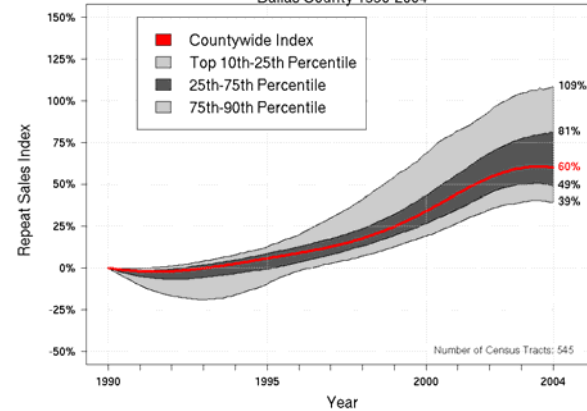
Variation of appreciation trends across Neighborhoods
Cuyahoga County 1990-2004



Variation of appreciation trends across Neighborhoods
King County 1990-2004



Variation of appreciation trends across Neighborhoods
Dallas County 1990-2004



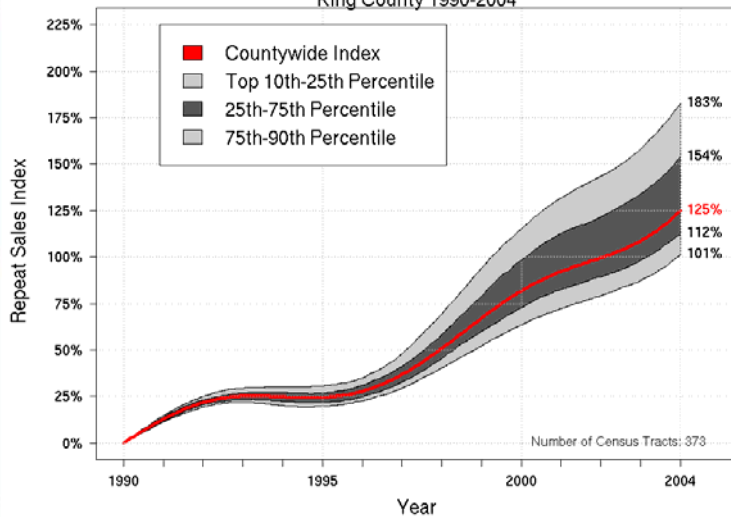
Most neighborhoods follow their region closely, but there are some exceptions

Neighborhoods and Regions

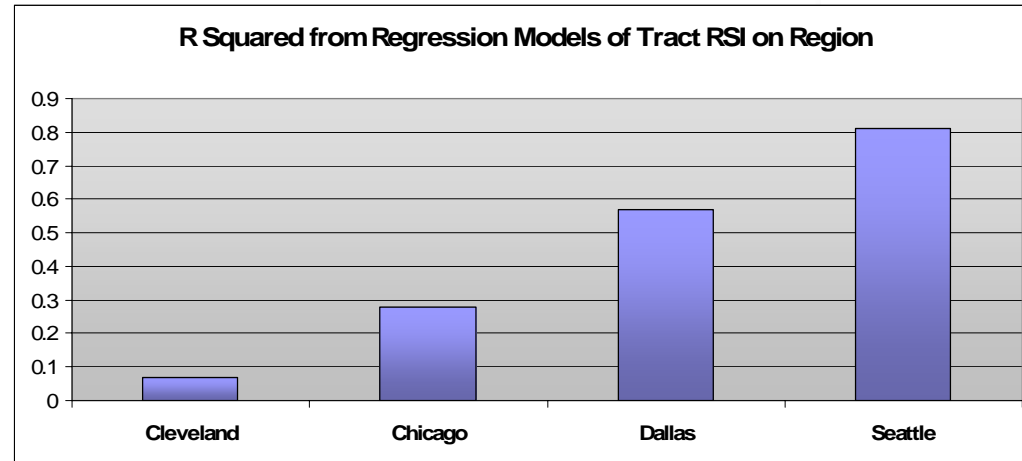
- Across Cities, 35% of Neighborhood Change is Accounted for by Regional Shifts
 - Regional shifts are more important in some regions than others

Variation of appreciation trends across Neighborhoods

King County 1990-2004



R Squared from Regression Models of Tract RSI on Region

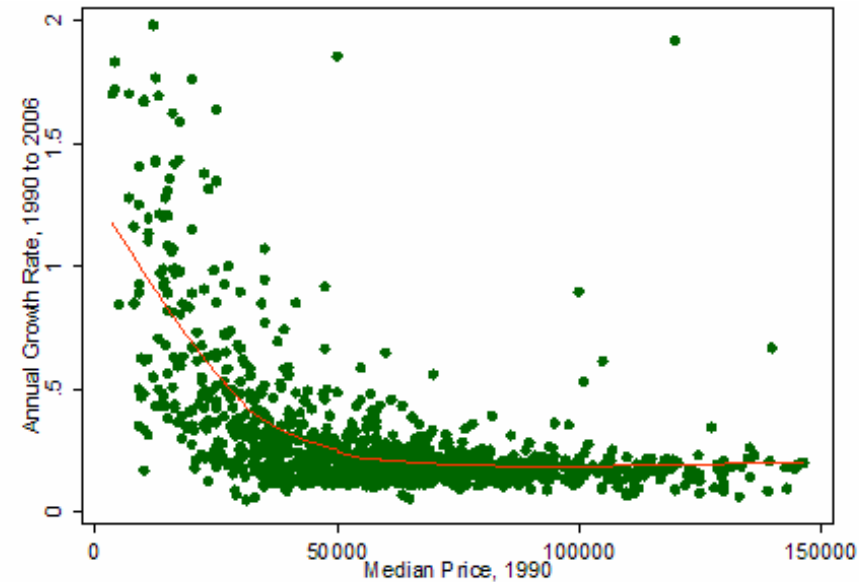


Localized movement in Cleveland; large regional impact in Seattle

Neighborhood Convergence

Sigma and Beta Convergence in Cook County, 1990-2006

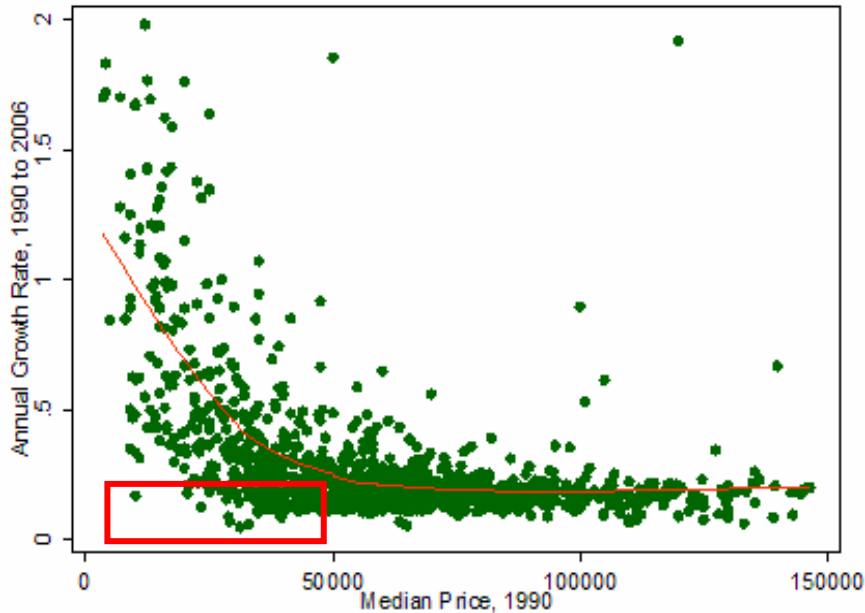
Variables	Obs.	Mean	Std. Err.	Std. Dev.	[95% Conf. Int.]	
ln_med_1990	1231	11.326	.01781	.62516	11.291	11.361
ln_med_2006	1307	12.419	.01414	.51126	12.391	12.447
Combined	2538	11.889	.01566	.78921	11.858	11.919
Ratio = sd (ln_median_y1990) / sd (ln_median_y2006)				f = 1.4952		
Ho: Ratio = 1		Degrees of freedom = 1230, 1306				
Ha: ratio < 1		Ha: ratio != 1		Ha: ratio > 1		
Pr(F < f) = 1.0000		2*Pr(F > f) = 0.0000		Pr(F > f) = 0.000		



The economic theory of convergence appears to apply at the neighborhood level as well, as neighborhoods tend to “catch up” with each other.

Neighborhood Convergence

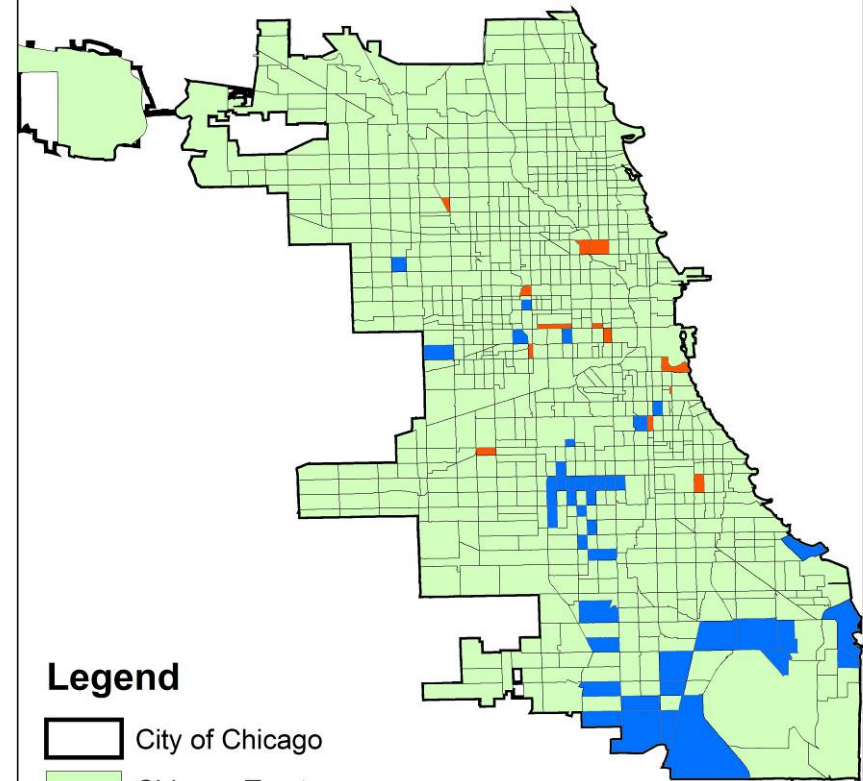
Beta Convergence in Cook County, 1990-2006



*Why Do Some
Neighborhoods Converge
while Others Don't?*

Neighborhoods Exhibiting "Divergence"

Chicago Tracts, 1990-2006



Legend

- City of Chicago
- Chicago Tracts
- Low Initial Values, Low Growth
- High Initial Values, High Growth

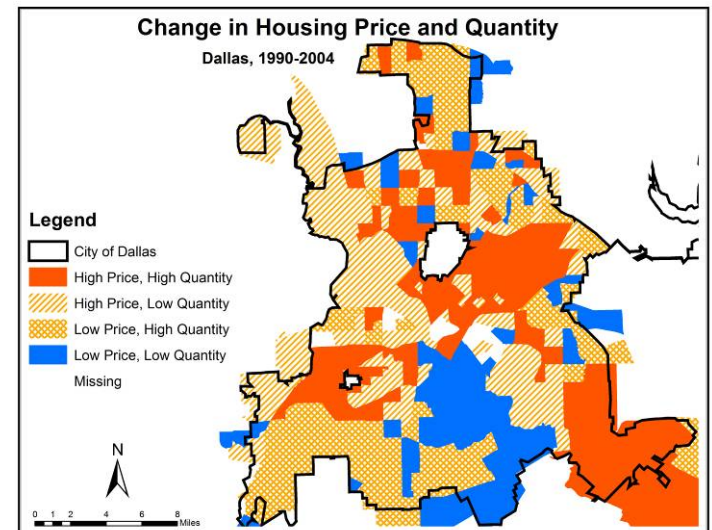
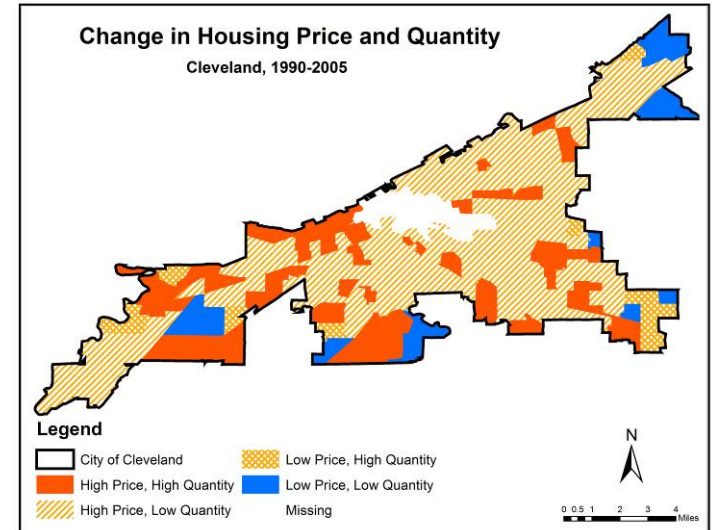
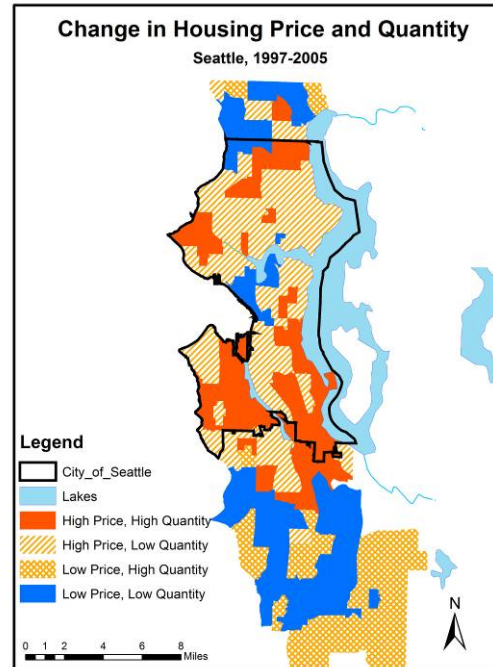
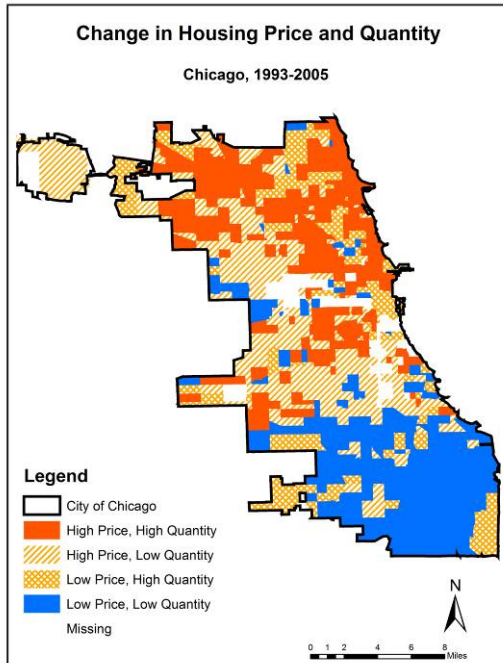


Neighborhood Change in 3D

- **Change in Demand for a Neighborhood will Result in:**
 - Change in Price
 - Change in Quantity
 - Change in Quality
- **The Combination of these Three Dimensions Gives Rise to Different Types of Neighborhood Change**



Combining the Dimensions

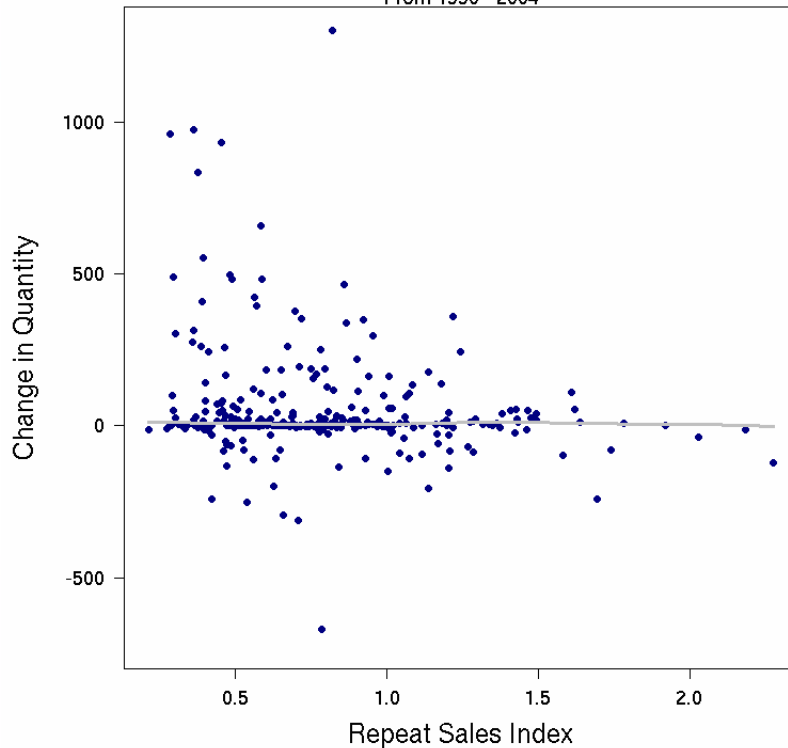


Why Do Some Poor Neighborhoods Show Explosive Growth While Others Remain "Cold"?

Relationship of Price and Quantity

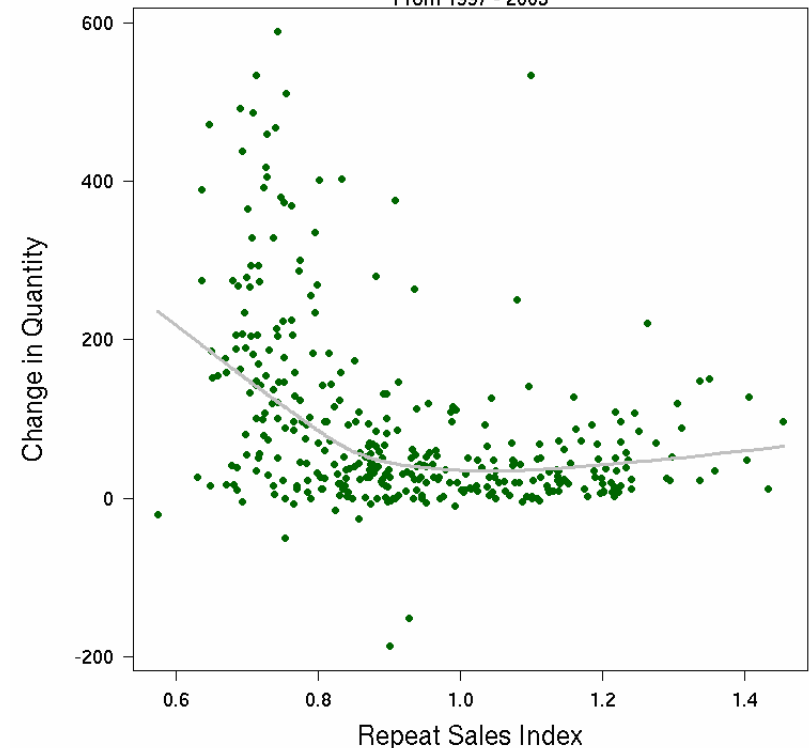
Appreciation vs. Change in Quantity in Dallas

From 1990 - 2004



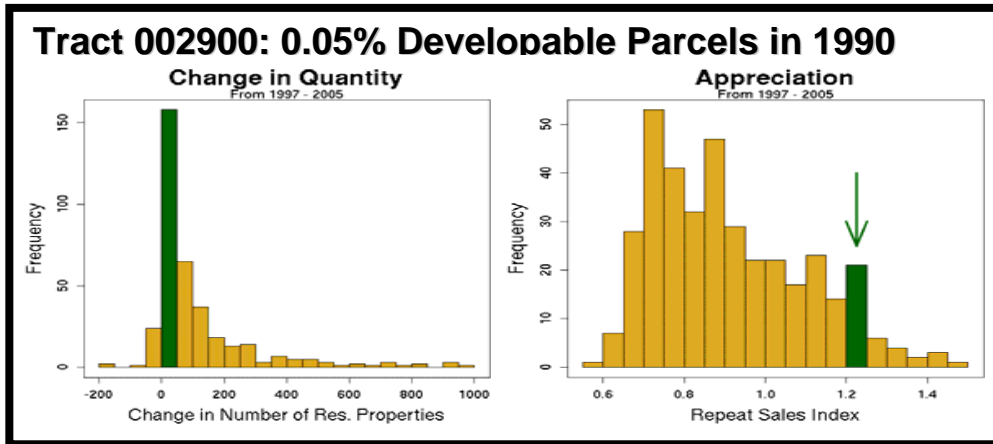
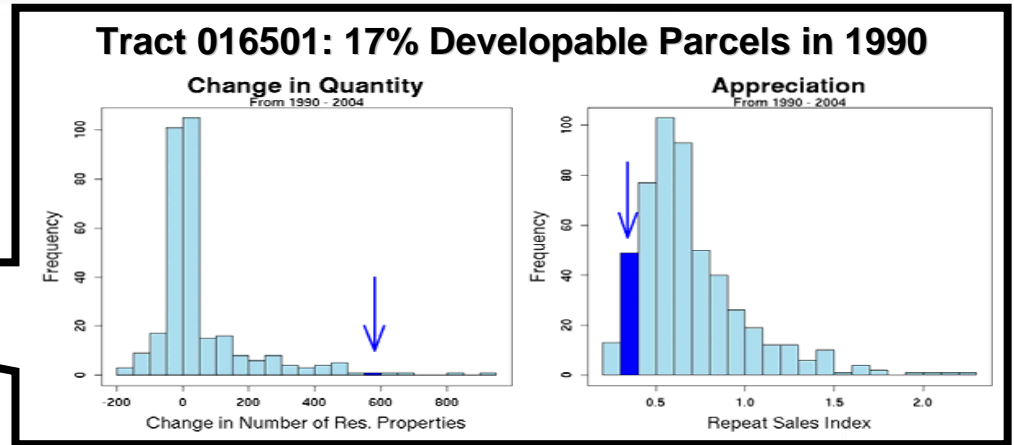
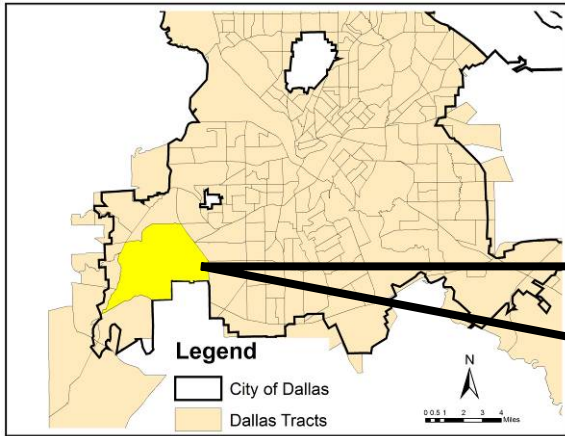
Appreciation vs. Change in Quantity in Seattle

From 1997 - 2005



Price and quantity are more negatively correlated in places where there are greater constraints on the supply of new housing units

New Development can Help Preserve Affordability



Neighborhoods with lower supply elasticity are at greater risk of displacement, as housing prices will increase faster than in areas where more housing units can easily be developed

Drivers Model and Data



Change in Amenities



Change in Demand for the Neighborhood



Change in Price

Change in Quantity

- **Physical:** Distance from CBD, vacancies, rehab activity, ...
- **Transportation:** Transit options, distance to jobs, ...
- **Consumption:** Retail, services, entertainment, ...
- **Public Services:** Quality of schools, police and fire, ...
- **Social Interactions:** Demographics, crime rates, social capital...



Drivers Analysis: Emerging Context and Story Lines

- Cities and urban neighborhoods are coming back
 - In this period of transition, the drivers of neighborhood change are evolving
- Neighborhood change occurs primarily through mobility
- Density matters
- Race is still a factor
- Neighborhood spillovers are important
- Context matters (starting point, type, ...)

