

Leveraging *Local* Data Sources

May 20, 2016

Sallie Keller, Director

Data characterizes the world and the ways in which communication occurs between its individuals

Infrastructure



- Condition
- Operations
- Resilience
- Sustainability

Environment



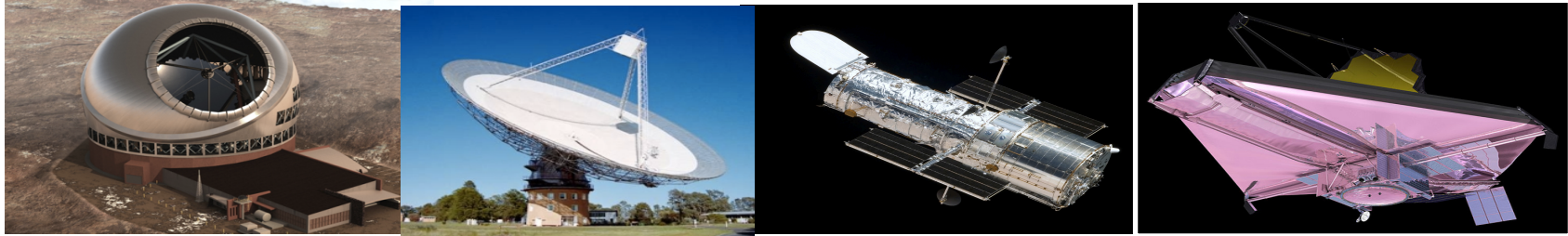
- Climate
- Pollution
- Noise
- Flora/ Fauna

People



- Relationships
- Location
- Economic Condition
- Communication
- Activities
- Health

The New Lens for Social Behavioral Observing



- Data collected faster, while individuals are in the act of behaving in real life situation
- Adapt methods to make the best use of these data
- New data streams produce new discoveries but should not be allowed to degrade the scientific approach

We are in the middle of an *ALL data* revolution

Designed Data Collections

Statistically-designed and intentional observational data collections



Administrative Data

Data collected for the administration of an organization or program



Opportunity Data

Data generated as we move through our daily paces



Procedural Data

Data derived from policies and procedures



What about data quality?

Traditional Approach:

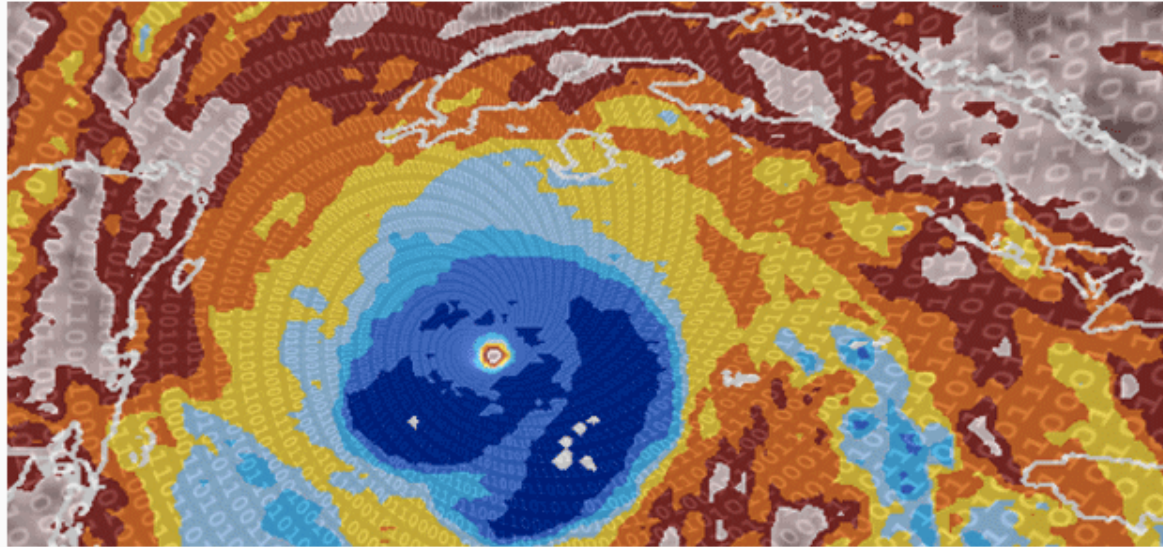
- Control over measurement processes
- Control over collection processes
 - Optimization
 - information maximization
- Clear and controlled ownership



Evolution of data quality by discipline

Discipline	Contributions to Data Quality
Physical and Biological Sciences	<ul style="list-style-type: none">• Experimental methods• Data repositories/portals• Reproducibility and replication
Engineering, IT, Business	<ul style="list-style-type: none">• Pareto Principle (80 - 20 rule)• Fitness-for-use• Total Data Quality Management (TDQM)• Data management• Standards
Social and Behavioral Sciences	<ul style="list-style-type: none">• Total Survey Error• Randomized Control Trials, Observational Studies, and Natural Experiments
Statistics	<ul style="list-style-type: none">• Decision theoretic approach• Statistical methods• Adoption of definitions, methods, and approaches by official statistics

In this data revolution, we have lost control!



5 keys to getting big data under control

By Paul McCloskey

May 20, 2013

And then there is privacy!

Does Big Data Change the Privacy Landscape: A Review of the Issues

Sallie Ann Keller, Stephanie
Shipp, and Aaron Schroeder

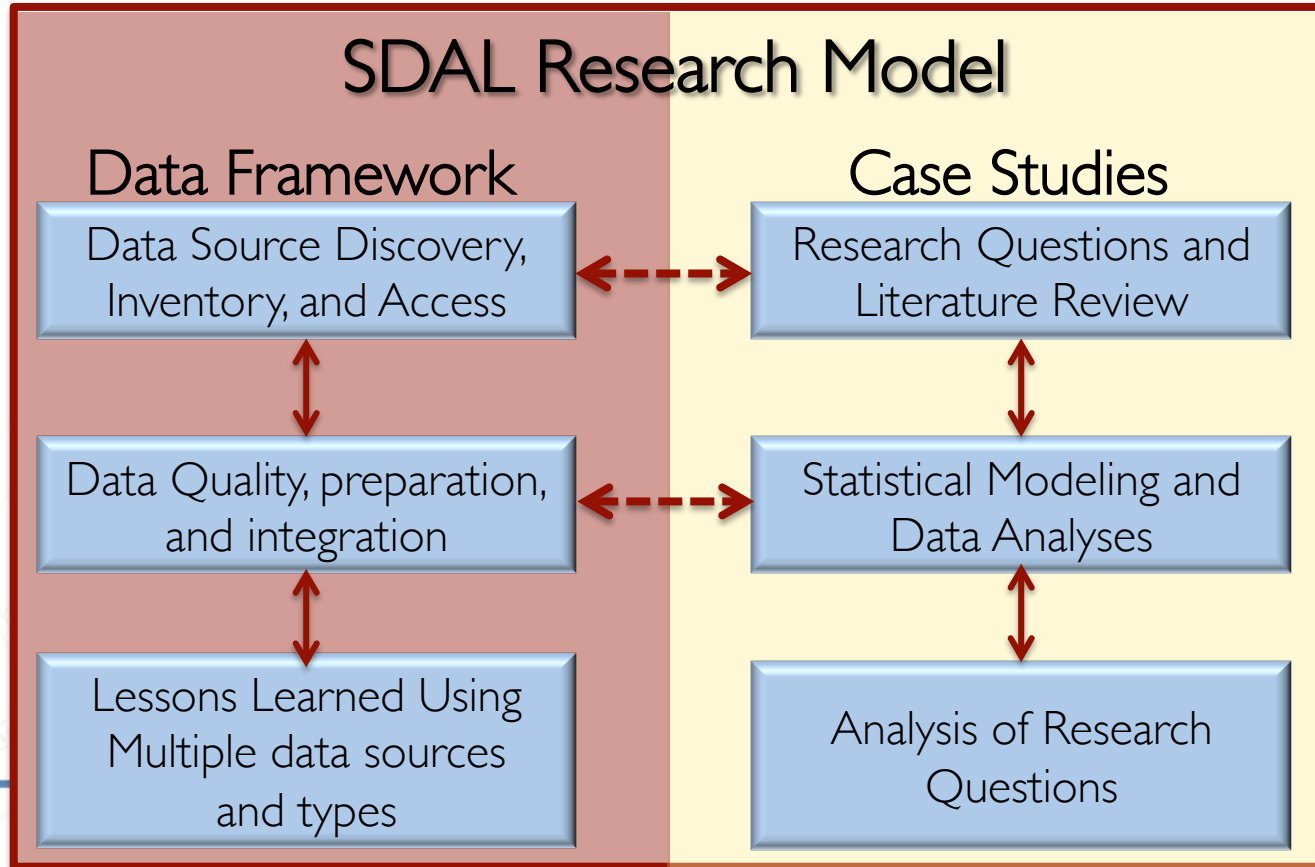


ANNUAL REVIEW OF STATISTICS AND ITS
APPLICATION VOLUME 3, 2016

Innovation data and measurement problems

- For what purpose?
 - Research, researcher access, policy, program evaluation, ...
- What dimension or aspect of innovation?
 - entrepreneurship, R&D, advances, productivity, productivity growth, technology, workforce, household innovators, automation, robotics, non-market diffusion, collective knowledge, ...
- Can we find local data surrogates?

A disciplined approach is needed to develop the theory and methods for using All the data



Data Framework

PROBLEM IDENTIFICATION: Relevant Issues and Working Hypotheses



LOCAL DATA
FEDERAL and STATE DATA
DESIGNED DATA
OPPORTUNISTIC DATA FLOWS

DATA
SOURCES:
Discovery
Inventory
Acquisition



DATA STORAGE



DATA PROFILING



DATA
PREPARATION



DATA LINKAGE



DATA
EXPLORATION



MODELING AND ANALYSES



FITNESS-FOR-USE ASSESSMENT

Data inventory process in-action

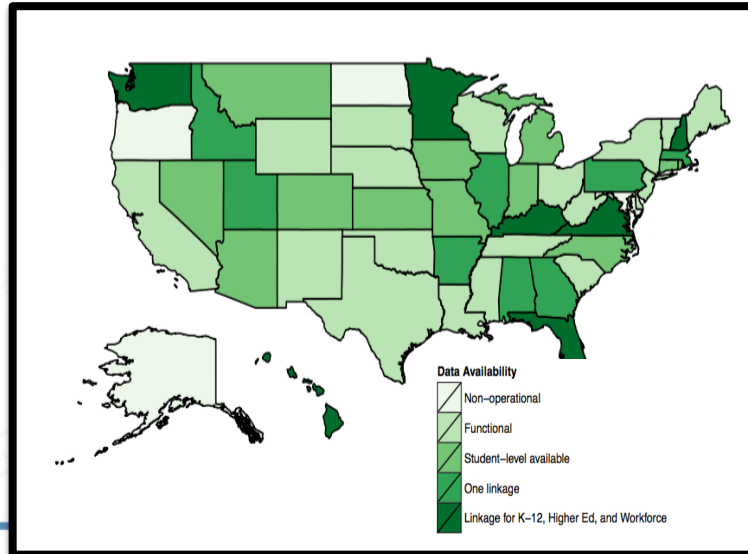
- Purpose of organization collecting the data
- Description of the data
 - content
 - Unit represented
 - Longitudinal or cross-sectional
 - Geographic coverage
 - Timing of collection and release
- Data collection method
- Metadata and provenance
- Other issues
 - Selectivity
 - Stability
 - Accessibility
 - Privacy and security
 - Research using data

Collaborative wiki – dynamic report appendix

Case study data sources that moved onto acquisition

Education – State Longitudinal Data Systems

- Washington
- North Carolina
- Virginia
- Kentucky
- Texas



Housing

- Local Count Real Estate Assessment Data
- Black Knight Financial Services
- CoreLogic
- MLS Data
- Local Crime Reports
- Location Inc.

Focus Briefly on Data Profiling

- Determination of quality of data and utility **to project at hand**
 - Data Quality
 - Data Structure
 - Metadata and Provenance
- Issues **are discovered but not fixed**, fix depends on research need



Local Data: Education Example

Exercising the data framework for the education data

- **Profiled variables**

- Student ID, district code, year, gender, race/ethnicity, grade, age, and other variables (e.g., limited English proficiency)
- Most variables were valid and consistent requiring very little cleaning

- **Transformed variables**

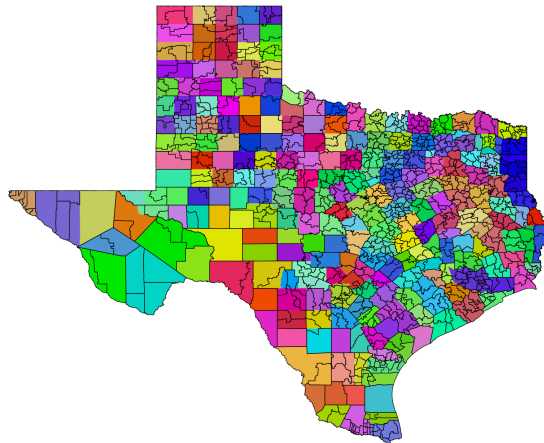
- Matched school districts with counties
- Calculated ages from birthdates (NC and KY)
- Texas enrollment estimates weighted to match the state level counts

- **Restructured variables**

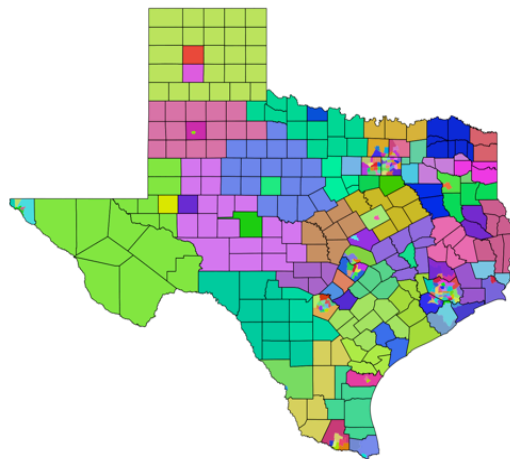
- Virginia data were restructured to create three main tables for race/ethnicity, by grade, gender by grade, and disadvantaged status by grade

Local data challenges with geographic alignment

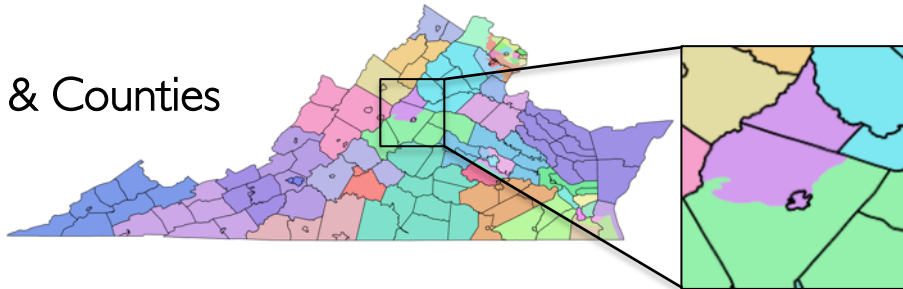
Counties & School districts



PUMAs & Counties



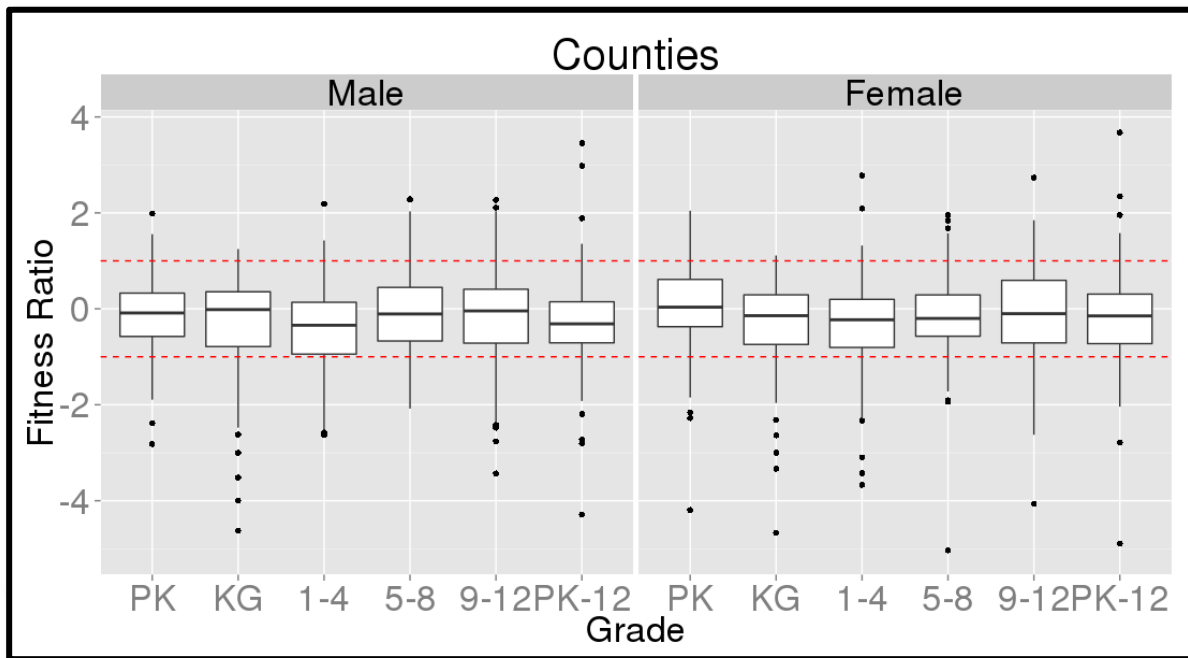
PUMAs & Counties



Education SLSD Alignment with ACS – Kentucky

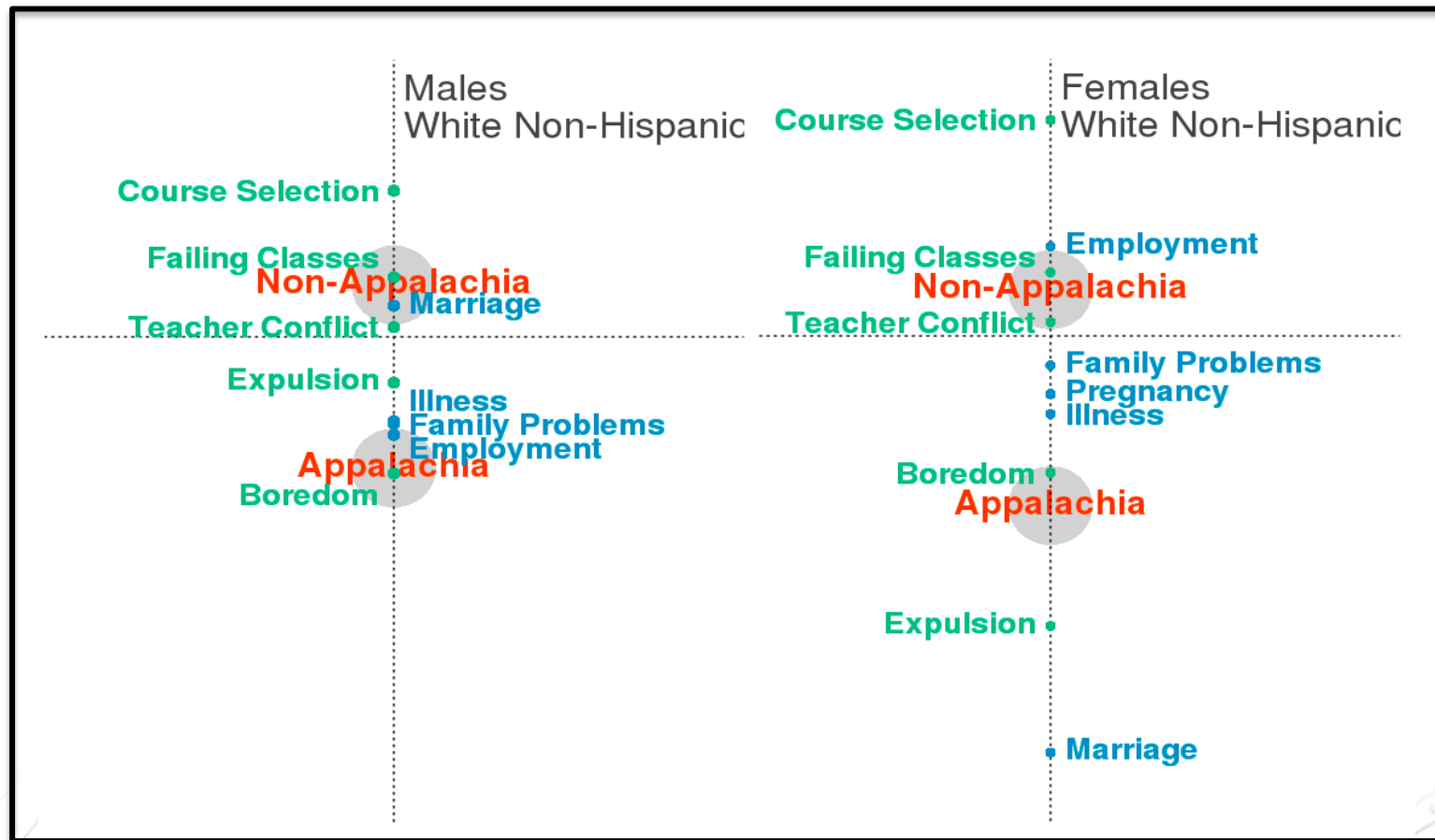
$$\text{Fitness Ratio} = \frac{\text{ACS Estimate} - \text{Local Data Estimate}}{\text{ACS 90\% Margin of Error}}$$

Public Enrollment at Kentucky State and County Levels, 2009-2013



Kentucky Public School Dropouts

Grades 9-12, KLDS 2009-2013



Local Data: Housing Example



Exercising the data framework for the housing data

- **Profiled variables**

- Need to align units of analysis, parcels versus housing units
- Reconcile land use codes through discussion with experts
- Remove non-residential parcels

- **Transformed variables**

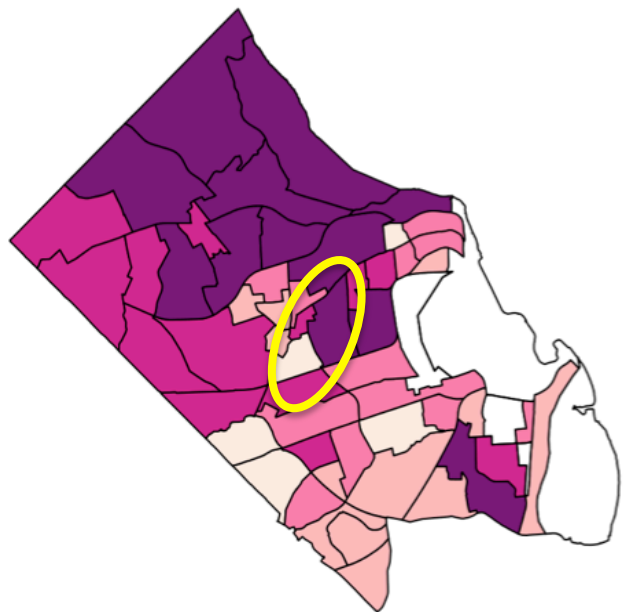
- Weight parcels by (estimated) number of units
- Adjusted inconsistent longitudinal data, e.g., number of units or year built

- **Restructured variables**

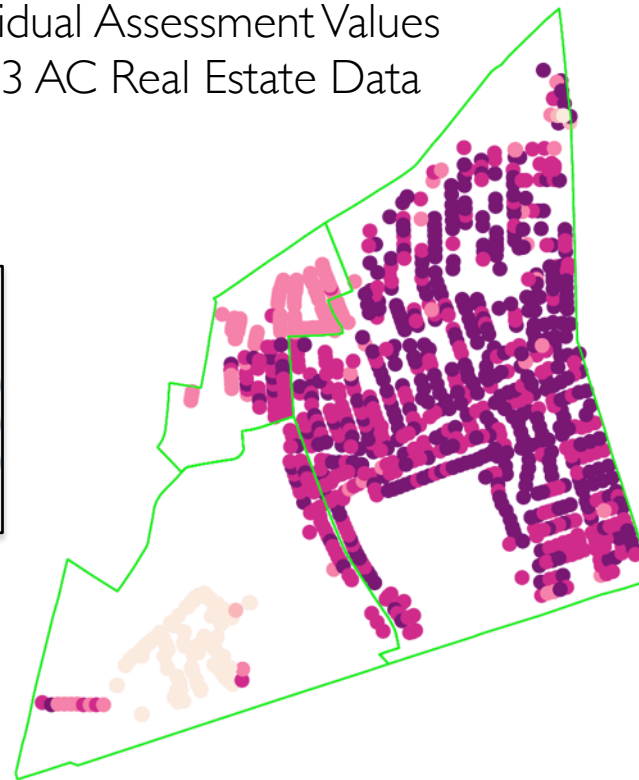
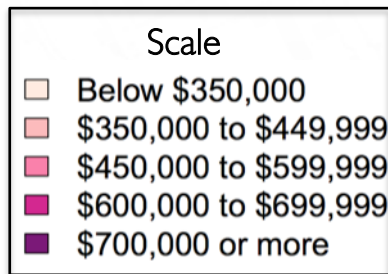
- Create a consistent set of geocodes per parcel

Characterizing neighborhoods

Median House Value
Owner Occupied Units
ACS 2009-2013 Estimate



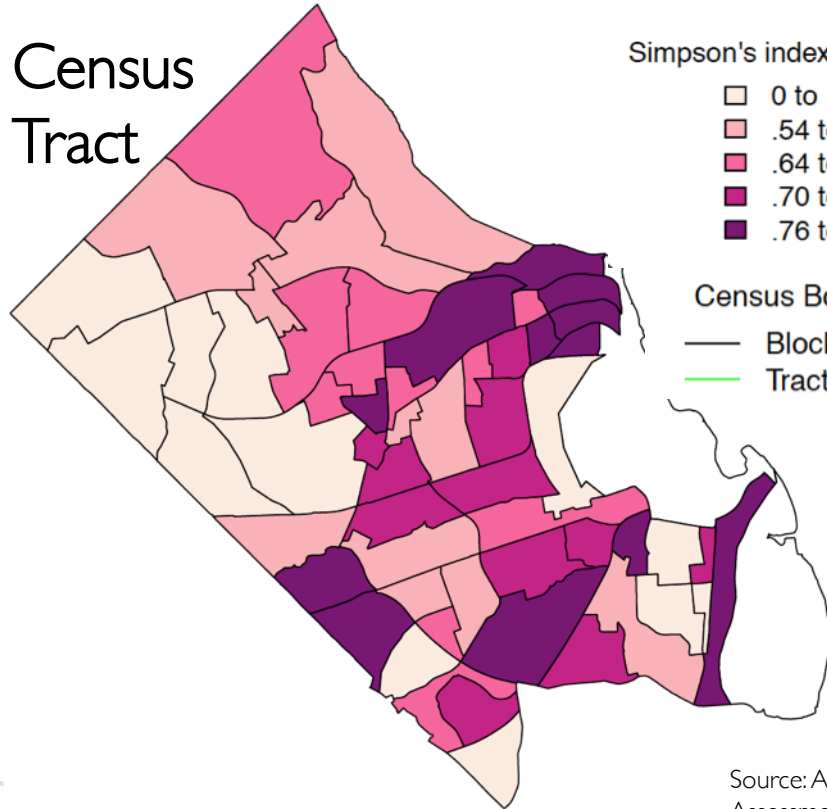
Individual Assessment Values
2013 AC Real Estate Data



Source: Arlington County Real Estate
Assessment, 2013; ACS 2009-2013

Simpson Index of Home Value Diversity

Census
Tract



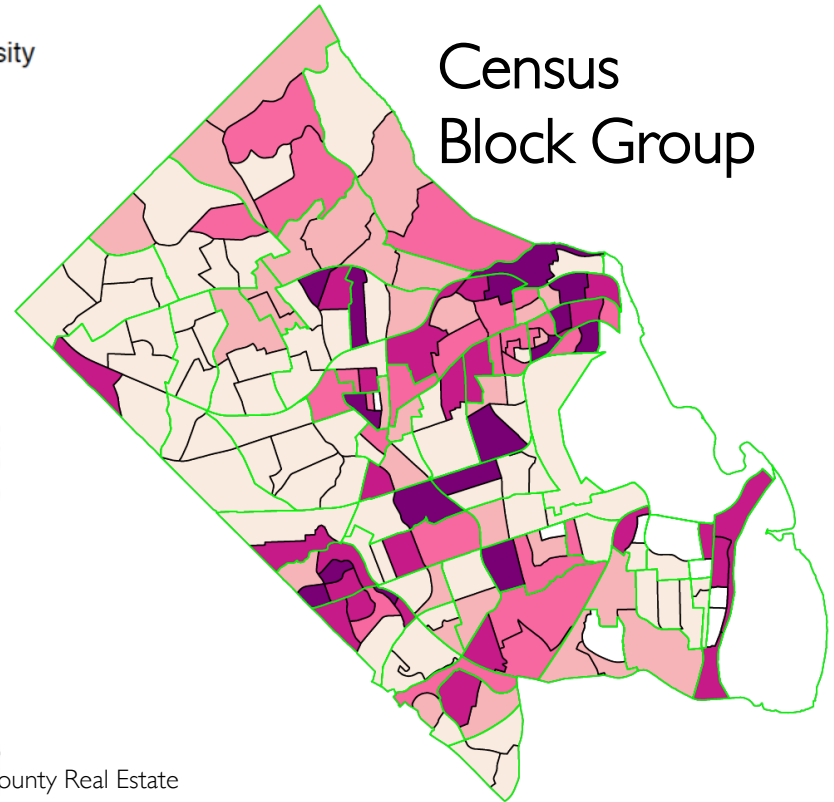
Simpson's index of diversity

- 0 to .53
- .54 to .63
- .64 to .59
- .70 to .75
- .76 to .86

Census Borders

- Block Group
- Tract

Census
Block Group



Source: Arlington County Real Estate
Assessment 2013

Concluding Remarks

- Use of **external data** – **no control over collection**, unlike federal statistics
- **Through lens of case studies** develop and test data framework
- Need a **disciplined, yet flexible and adaptable, data framework** to assess data quality and fitness-for-use that is dependent on use, e.g., Innovation measures

Proposed Criteria for Future Research

- Speed and regularity of acquiring data
- Flexibility in negotiating data agreements
- Capacity to implement Data Framework
 - e.g., Organizational structure and constraints, GIS, longitudinal discrepancies, statistical data integration, etc.
- Research needed to align and harmonize data sources
- Value proposition to sponsor