University of Massachusetts Boston

From the SelectedWorks of Michael P. Johnson

November 3, 2016

Decision Modeling for Housing and Community Development: ATM ethodology for Evidence-Based Urban and Regional Planning

Michael P Johnson, Jr.



Decision Modeling for Housing and Community Development: A Methodology for Evidence-Based Urban and Regional Planning

Michael P. Johnson

Department of Public Policy and Public Affairs and

Urban Planning and Community Development Master's Program

University of Massachusetts Boston

michael.johnson@umb.edu

56th Annual Conference of the Association of Collegiate Schools of Planning, Portland, OR, November 3, 2016

Motivation

- Analytic skills in urban planning and related programs typically provided through probability and statistics, geographic information systems and qualitative data analysis
- Problem-solving is, and should properly be seen as, driven by stakeholder needs, not available technologies
- In some contexts, and for some organizations, methods associated with 'analytics' or 'decision science' can provide unique insights.

How can data analytics and decision science provide useful guidance to planners engaged in local housing and community development beyond standard methods?

What are some different quantitatively-oriented new technologies and methods?

- Big data: collection of models, methods, tools and repositories designed to extract knowledge and insight from very large datasets (Kitchin 2014)
- Smart cities: technologies often based on real-time data to support more efficient and responsive provision of services in urban areas (Geertman et al. 2015)
- (Decision or data) Analytics: procedures and models intended to generate guidance regarding design and operation of organizations, systems and phenomena (Liberatore and Luo 2010):
 - Current characteristics (descriptive analytics)
 - Future values of important variables (predictive analytics)
 - Prescriptions , policy and guidelines (prescriptive analytics)

What types of problems in housing and community development may be well-suited for analytics?

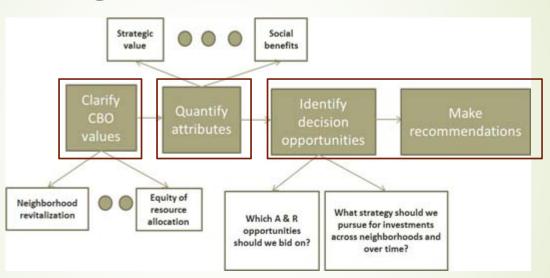
- Characteristics
 - Multi-stakeholder
 - Multi-objective
 - Support operational, tactical or strategic decisions
 - Rooted in values
 - Accommodate multiple methodological approaches
- Examples
 - Subsidized and affordable housing (Johnson 2007; 2005)
 - Foreclosure response (Johnson et al. 2016)
 - Vacant land management (Johnson, Hollander and Whiteman 2015; Johnson, Hollander and Hallulli 2014)

There are other sources of insight related to data analytics for housing and community development...

- Mallach's framework for strategic change (2008)
- Foreclosure recovery timeline (National Housing Center 2013)
- Resources for organization capacity-building and data analysis (e.g. NeighborWorks; PolicyMap)
- Evidence-based local investment strategies for community development (Federal Reserve Bank of San Francisco & Low Income Investment Fund 2012)

However, when we don't know what metrics to use, what tradeoffs to consider, which alternatives to choose, which policies to follow, how to incorporate uncertainty, analytics can generate new and useful solutions

How can we apply data and decision analytics to foreclosed housing response?



- Values analysis: identify strategy/intervention objectives and decision attributes
- Data analytics: quantify attributes; refine understanding of problem context
- Decision modeling: Choose most-preferred decision alternative; design optimal planning strategy

Values analysis: Generate values structures to evaluate strategy alternatives General goal:

determine

relationship

alternatives,

metrics and

preferences

(Keeney 1992):

identify causal

links between

fundamental

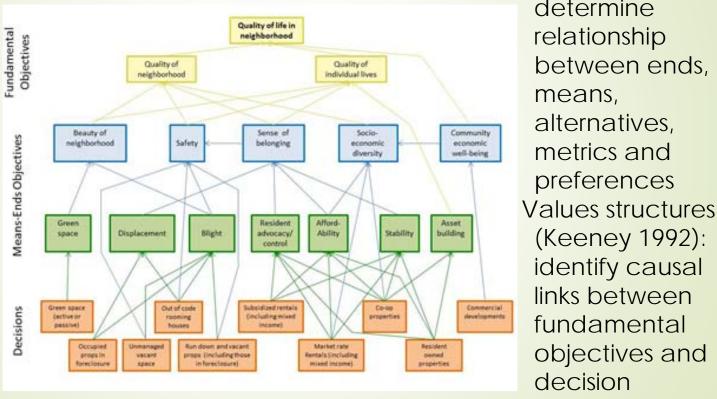
decision

alternatives

objectives and

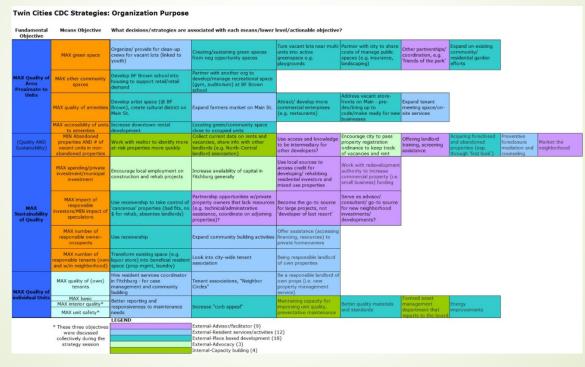
means,

between ends.



Source: Johnson, et al. (2016)

Values analysis: Classify action items according to means objectives



'Strategy
tables' can
support
decisionmaking
without
need to
quantify
preferences
or define
tradeoffs

Source: Johnson, et al. (2016)

Data analytics: Provide evidencebased context for specific interventions

	FORECLOSURE IMPACT RISK							
MARKET STRENGTH	C. Actual high foreclosure density	B. High risk of high foreclosure density	A. Low risk of high foreclosure density					
1. Strong	Facilitate rapid sales to sustainable owners, low/no subsidy	Lower cost effort to prevent foreclosures and vacancies, low/no subsidy	Lower priority					
2. Intermediate	High payoff/priority, rehab and rapid sale to sustainable owners, target subsidies, neighborhood maintenance	High payoff/priority, prevent foreclosures and vacancies, emphasize neighborhood maintenance	Lower priority but watch carefully, head-off emerging problems early					
3. Weak	More emphasis on securing/demolishing, land banking to hold until market rebound	Lower cost effort to prevent foreclosures and vacancies	Lower priority but watch carefully, head-off emerging problems early					

Source: foreclosure-response.org (2013)

This foreclosure response typology can help CDCs classify interventions – but which are most relevant to which neighborhoods, according to what thresholds?

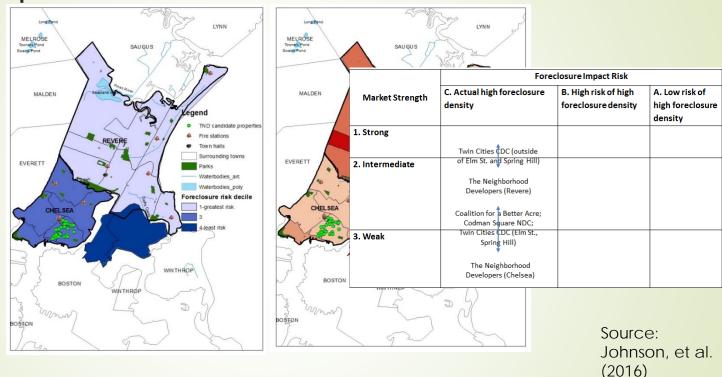
Data analytics: Quantify criteria and measure impact

Fitchburg/Leominster, MA													
Strength	Weakest> Strongest	10	0	0	0	0	0	0	0	0	0	0	
		9	0	0	0	0	0	0	0	0	0	0	
		8	0	0	0	0	0	0	0	0	0	0	
		7	1	0	0	0	0	0	0	0	0	0	
ř		6	1		0	0	0	0	0	0	0	0	
Market S		5	1	1	0	0	0	0	0	0	0	0	
		4	0	2	0	0	0	0	0	0	0	0	
		3	1	2	0	0	0	0	0	0	0	0	
		2		0	0	0	0	0	0	0	0	0	
		1		0	0	0	0	0	0	0	0	0	
			1	2	3	4	5	6	7	8	9	10	
		Highest>> Lowest											
Foreclosure Risk													

Source: Johnson, et al. (2016)

For each of four partner communities, classified each Census tract according to foreclosure risk and market strength to determine what types of responses might be most relevant

Data analytics: Identify areas for specific interventions



By overlaying foreclosure incidents with neighborhood characteristics, we can determine what types of interventions may be most suitable in which communities

Decision modeling: Design interventions for specific communities

Model jointly optimizes estimated strategic value and property value impact associated with foreclosure acquisitions

Optimize
$$\{S(\mathbf{x}) = \sum_{i=1}^{n} S_i \cdot x_i; P(\mathbf{x}) = \sum_{i=1}^{n} P_i \cdot x_i\}$$
 — Jointly optimize social objectives

$$\sum_{i=1}^{n} C_i \cdot x_l \leq B$$

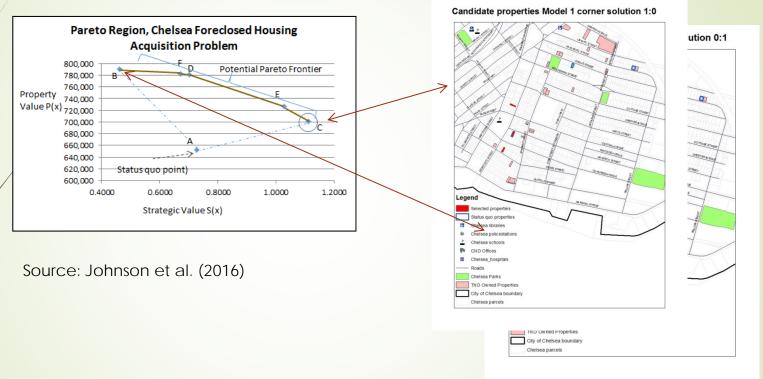
 Limit expenditures to budget available

- or -

$$\sum_{i=1}^{n} x_i = N$$
 Acquire only a given number of properties

 $x_i \in \{0, 1\}, i = 1, ..., n$

Decision modeling: Generate alternative acquisition strategies

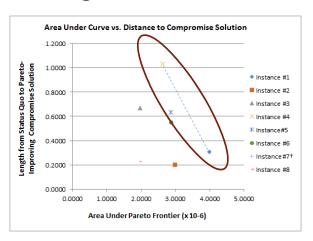


Spatial specificity of solutions is valuable – but which modeling strategy is preferred? Which strategic value metric should we use? Which acquisition strategy should the CDC pursue?

Decision modeling: Choose mostpreferred modeling approach and intervention strategy

Number-acquired

Budget constraint



Source: Johnson, et al. (2016)

Area Under Curve vs. Distance to Compromise Solution†

1.4000

1.2000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

1.0000

By systematically varying model formulations and parameters, we can determine which approach is likely to maximize social value, and which specific intervention the CDC may choose

Additional applications of data and decision analytics

- Vacant land management
 - Identify alternative uses of currently unoccupied land that maximize multiple social impact metrics
 - Neighborhood-level strategies (Johnson, Hollander and Hallulli (2014)
 - Parcel-level strategies (Johnson, Hollander and Davenport Whiteman (2015)
- Gentrification and displacement
 - Identify and prioritize local responses based on a values analysis, emphasizing local knowledge and communitybuilding rather than quantitative analysis
- Subsidized and affordable housing provision
 - Choose vacant parcels for affordable housing construction that balance accessibility to employment and perceived impacts on nearby communities

Conclusion and next steps

- Key findings:
 - Values analysis enables stakeholders to clarify objectives and goals to generate performance metrics and strategy alternatives that represent decision opportunities
 - Data analytics can quantify decision opportunities and performance metrics to complement experience and intuition
 - Decision modeling can generate specific yet flexible guidance regarding multiple aspects of foreclosure response: selection of housing units for potential acquisition; bidding strategies; neighborhood-level redevelopment under uncertainty
- Related research opportunities
 - Develop theoretical framework to help determine conditions under which decision modeling is likely to add the greatest value
 - Design and implement truly community-based and communityengaged decision modeling applications
 - Formally evaluate data and decision analytics-based solutions as compared to current or conventional practice

References

- Geertman, S., Ferreira, J., Jr., Goodspeed, R. and J. Stillwell (Eds.) 2015. Planning Support Systems and Smart Cities.
 Cham, Switzerland: Springer.
- Federal Reserve Bank of San Francisco & Low Income Investment Fund 2012. Investing in What Works for America's Communities: Essays on People, Place and Purpose. San Francisco
- Johnson, M.P. 2007. Planning Models for the Provision of Affordable Housing. Environment and Planning B: Planning and Design 34(3): 501 – 523.
- Johnson, M.P. 2005. Spatial Decision Support for Assisted Housing Mobility Counseling. *Decision Support Systems* 41(1): 296 312.
- Johnson, M.P., Hollander, J. and E. Davenport Whiteman. 2015. "Data and Analytics for Neighborhood Development: Smart Shrinkage Decision Modeling in Baltimore, Maryland". In (Stan Geertman, Joe Ferreira, Robert Goodspeed and John Stillwell, Eds.) Planning Support Systems and Smart Cities. Switzerland: Springer, p. 61 76.
- Johnson, M.P., Hollander, J. and A. Hallulli. 2014. Maintain, Demolish, Re-purpose: Policy Design for Vacant Land Management using Decision Models. *Cities: Special Issue: Vacant Land: The New Urban Green* 40:151 162.
- Johnson, M.P., Keisler, J., Solak, S., Turcotte, D., Bayram, A. and R.B. Drew. 2016. *Decision Science for Housing and Community Development: Localized and Evidence-Based Responses to Distressed Housing and Blighted Communities*. New York: John Wiley & Sons, Inc.
- Keeney, R.L. 1992. Value Focused Thinking: A Path to Creative Decision Making. Cambridge, MA: Harvard University Press.
- Kitchin, R. 2014. The Data Revolution: Big Data, Open Data, Data Infrastructures and Their Consequences. London: Sage Publications, Ltd.
- Liberatore, M.J. and W. Luo. 2010. The Analytics Movement: Implications for Operations Research. *Interfaces* **40**(4): 313 324.
- Mallach, A. 2008. Managing Neighborhood Change: A Framework for Sustainable and Equitable Revitalization. Montclair, NJ: National Housing Institute. Web: http://www.nhi.org/pdf/ManagingNeighborhoodChange.pdf.
- Neighborworks. http://www.neighborhworks.org
- PolicyMap. http://www.policymap.org.