



**DIFFERENCES IN ACCESSIBILITY
SPATIAL MEASURES:
IMPLICATIONS FOR COMMUNITY
STRATEGY**

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PRESENTER DISCLOSURES

Lara Jaskiewicz

(1) The following personal financial relationships with commercial interests relevant to this presentation existed during the past 12 months:
No relationships to disclose

PURPOSE OF THE STUDY

- Explore whether different measures of spatial accessibility identify the same areas of high and low access
- Explore whether population characteristics predict supermarket accessibility similarly across spatial measures

ACCESSIBILITY

Penchansky & Thomas (1981) framework

- Availability
- Accessibility
- Affordability
- Acceptability
- Accommodation

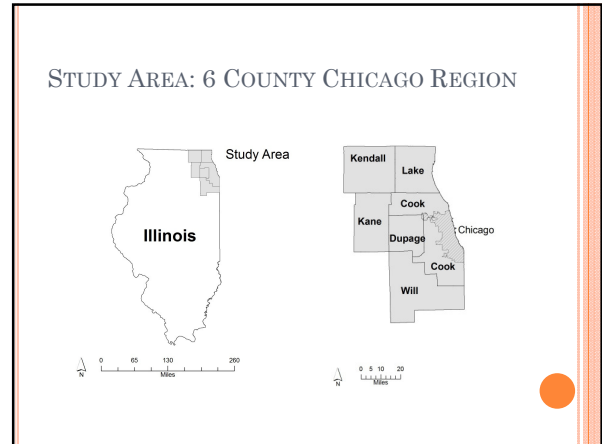
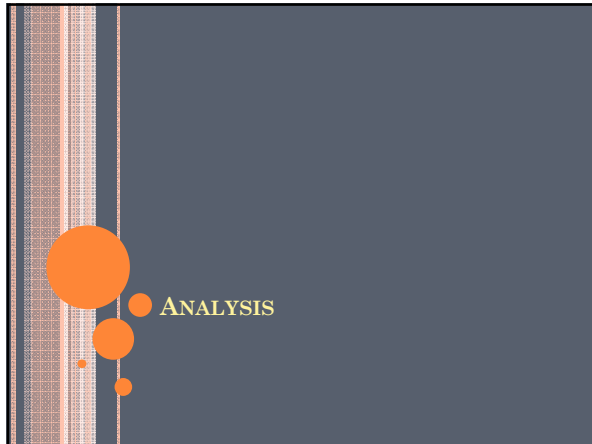
This study—availability and accessibility

ACCESSIBILITY

- Potential vs. realized
- Different measures
 - Cumulative opportunities
 - Gravity models
 - Utility models
 - *Little comparison of measures in literature*

PREVIOUS SUPERMARKET ACCESS STUDIES

- Container measure most common
- Low-income and low-income Black usually predictive of less access
- Few studies looked at other races or ethnicities
- Limitations
 - Cumulative opportunities measures
 - Single measures
 - Modifiable area unit problem
 - Lack of independence
 - Spatial clustering



STUDY DATA: SUPERMARKET LOCATIONS

- Industry datasets
- Visual confirmation & additions – 2007
- Population/store
 - 10,261 in Chicago
 - 7,029 in McHenry County
- Supermarket = >\$2 mill in annual sales

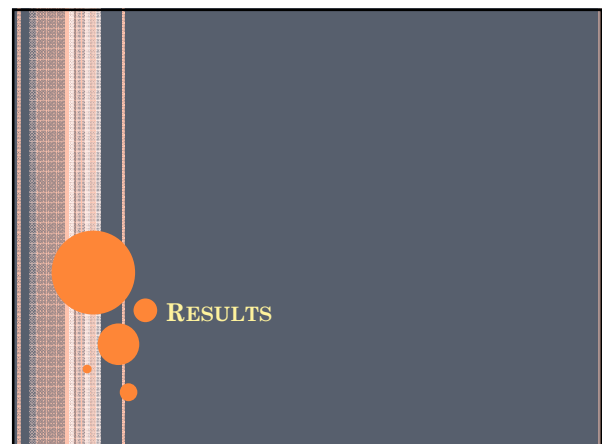
Store Type	#	%
Full Service Chain	258	31.2
Discount	118	14.3
Independent	355	43.0
Specialty	27	3.3
Supercenter	28	3.4
Wholesale and Warehouse	40	4.8
Total	826	100

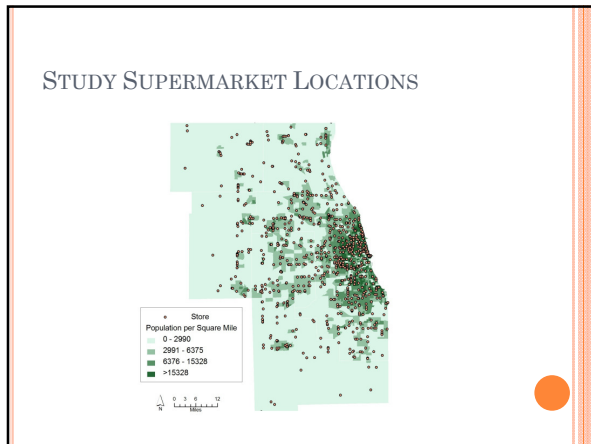
STUDY DATA: POPULATION

- 2000 US Census
- Populated units
 - Block groups – 5,779
 - Census tracts – 1,826
 - Zip codes – 350
- Variables
 - Population density
 - Race/ethnicity
 - Education
 - Median & aggregate household income
 - Family size
 - Home ownership & value

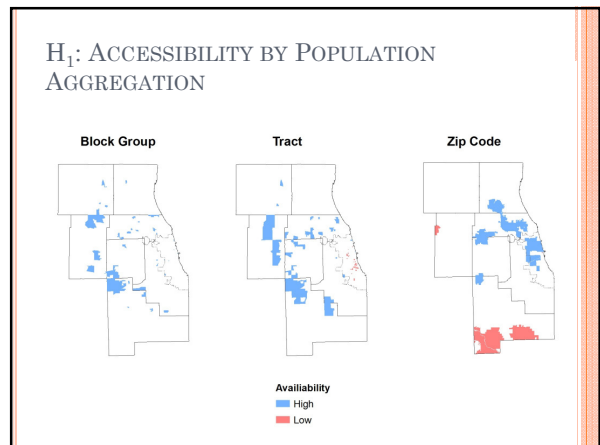
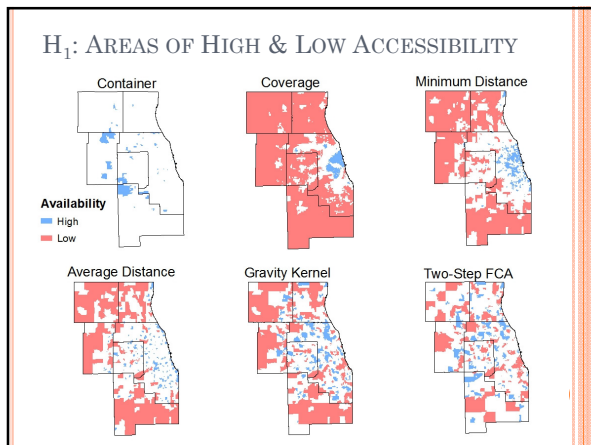
ACCESSIBILITY MEASURES

- 6 Measures Studied
 - Cumulative Opportunities
 - Container: # stores/area (block group, census tract, zip code)
 - Coverage: # stores w/i 2 miles of block group centroid
 - Minimum Distance: distance to closest store
 - Average Distance: avg. distance to all stores in 2 mile radius
 - Gravity Models
 - Gravity kernel: population density divided by density of stores w/i 2 miles
 - Two-step floating catchment area: facilities w/i 2 miles weighted by catchment population
 - Network distance & population weighted centroids
- Software
 - ArcGIS 9.2, Network Analyst & Spatial Analyst Extensions
 - GeoDa 0.9.8.8
 - SPSS 15.0





- ### H₁: DIFFERENT MEASURES IDENTIFY DISSIMILAR AREAS OF HIGH AND LOW ACCESS
- | Correlations | Moran's <i>I</i> (spatial clustering) |
|--|--|
| <ul style="list-style-type: none"> ○ Container – lowest <ul style="list-style-type: none"> • 0.024 - 0.270 ○ All others higher ○ Highest (>0.4) <ul style="list-style-type: none"> • Cov-MinD = 0.443 • MinD-AvgD=0.688 • Kernel-FCA=0.658 | <ul style="list-style-type: none"> ○ Global <ul style="list-style-type: none"> • Container very low • Others >0.4 • Coverage = 0.94 ○ Local <ul style="list-style-type: none"> • Correlations similar but much weaker • Two relationships >0.2 <ul style="list-style-type: none"> ○ MinD-AvgD = 0.552 ○ Kernel-FCA = 0.521 |



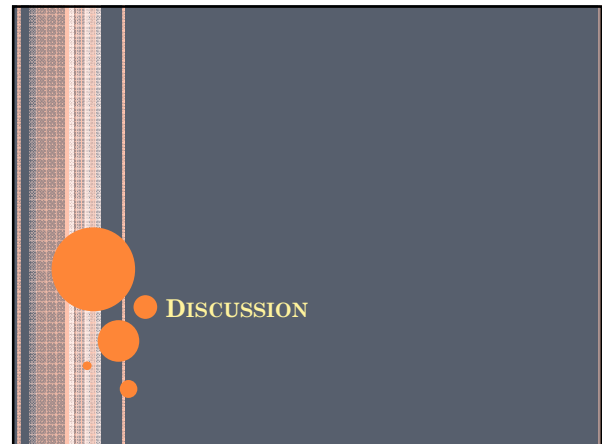
- ### H₁: SIMILARITY OF CLUSTERING
- Agreement of high/low between measures
 - Container: no low accessibility clustering and smallest number of high clustering areas
 - Min Distance: most high clustering areas – 1393
 - Urban – small geographic block group size
 - Coverage, Kernel, FCA: # low > # high
 - Disagreement of high/low between measures
 - Container disagreed more with Coverage than agreed
 - Min Distance & Kernel disagreed same # as agreed
 - Min Distance disagreed more with FCA than agreed
 - Strong disagreement b/w Avg Distance and gravity measures

- ### H₂: POPULATION VARIABLES DIFFER BETWEEN SUPERMARKET ACCESSIBILITY MEASURES
- Correlation
 - Stepwise OLS
 - Spatial dependency and ML modeling
 - Model fit – Akaike Information Criterion
 - Variable relationships
 - R^2

H₂: FINAL MODEL VARIABLE RELATIONSHIPS – ALL LARGE STORES

Measure	Pop Dens	% Black	% Hisp	Med HH Inc	Agg HH Inc	% HH4+	Med HomVel	% Pov	Model Type	Adj. R ²
Container Zip Code			+		+				OLS	.394
Container Tract	-	-		-	+				Spatial Error	.120
Container Block Group	-	-			+	-			OLS	.055
Coverage	+		+		-	-	+	-	Spatial Lag	.461
Minimum Distance	+	+	+	-	-	-	+	-	Spatial Lag	.157
Average Distance	+		+	-	-	-			Spatial Lag	.097
Gravity Kernel	-	-					-		Spatial Lag	.076
2-Step Floating Catchment	-	-			-		-	-	Spatial Lag	.072

All p-values < 0.0001. Continued spatial dependency significant at < 0.0001



- ### DISCUSSION
- Know your community
 - Measures are different
 - Different high/low accessibility areas
 - Different relationships with population variables
 - Cannot combine for meta-analysis
 - Match the measure to the question
 - Use more than one measure

- ### DISCUSSION
- Modifiable areal unit problem is important
 - Spatial dependency, independent variables
 - Choose the smallest meaningful population unit
 - Spatial dependency plays a role
 - Regression modeling—maximum likelihood
 - Additional variables
 - Store location explained by more than local population
 - Low R²
 - Include other explanatory factors, e.g., taxes, space, etc.

- ### STUDY LIMITATIONS
- Cross-sectional
 - Ecological
 - Within/between-area variability: modifiable areal unit problem; travel distance vs. travel time
 - Unmeasured confounding: non-population store location variables; shopper preference and behavior
 - Effect modification: variables may not be discrete
 - Contextual effects: non-spatial accessibility factors
 - Measurement error: missing stores; alternate food sources
 - Assumption of car use for shopping

