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Abstract

Social scientists studying the disadvantages of poor urban neighborhoods have tended to focus on the quality of publicly provided amenities, such as public schools and crime rates, or negative peer effects (Case & Katz, 1991; Cutler, Glaeser, & Vigdor, 1999; Jargowsky, 2003; Massey & Denton, 1993; Wilson, 1987). However, the quantity and quality of local private amenities, such as grocery stores, restaurants, banking facilities, and other retail services, can also have important quality-of-life implications for neighborhood residents. The “consumer city” literature suggests that attractive and abundant retail services affect a city’s ability to attract and retain high-skilled residents; by extension, the quality of neighborhood retail may affect the neighborhood’s growth prospects (Glaeser & Gottlieb, 2006; Glaeser, Kolko, & Saiz, 2001). A smaller empirical literature has shown that low-income and minority neighborhoods are typically less well served by certain types of retail and household services (Alwitt & Donley, 1997; Carr & Schuetz, 2001; Helling & Sawicki, 2003; Lewis et al., 2005; Pearce, Blakely, Witten, & Bartie, 2007; Powell, Slater, Mirtcheva, Bao, & Chaloupka, 2007; Zenk, 2005). In general, a smaller

Introduction

Social scientists studying the disadvantages of poor urban neighborhoods have tended to focus on the quality of publicly provided amenities, such as public schools and crime rates, or negative peer effects (Case & Katz, 1991; Cutler, Glaeser, & Vigdor, 1999; Jargowsky, 2003; Massey & Denton, 1993; Wilson, 1987). However, the quantity and quality of local private amenities, such as grocery stores, restaurants, banking facilities, and other retail services, can also have important quality-of-life implications for neighborhood residents. The “consumer city” literature suggests that attractive and abundant retail services affect a city’s ability to attract and retain high-skilled residents; by extension, the quality of neighborhood retail may affect the neighborhood’s growth prospects (Glaeser & Gottlieb, 2006; Glaeser, Kolko, & Saiz, 2001). A smaller empirical literature has shown that low-income and minority neighborhoods are typically less well served by certain types of retail and household services (Alwitt & Donley, 1997; Carr & Schuetz, 2001; Helling & Sawicki, 2003; Lewis et al., 2005; Pearce, Blakely, Witten, & Bartie, 2007; Powell, Slater, Mirtcheva, Bao, & Chaloupka, 2007; Zenk, 2005). In general, a smaller

number of retail outlets implies a more limited choice, and an apparent lack of competition has led some researchers to argue that “the poor pay more” for many basic goods and services ([Caplovitz, 1967](#); [Hayes, 2000](#); [Kaufman, MacDonald, Lutz, & Smallwood, 1997](#)). Therefore, understanding the extent and reasons behind differences in the amount and composition of neighborhood commercial activity is an important area for research.

To date, there has been little large-scale empirical work looking at disparities in retail services across neighborhoods of varying economic and demographic compositions. Moreover, there has been no work looking at the change in neighborhood retail services over time and how these changes correspond with economic and demographic changes in the local population. Therefore, before we can explain the causes behind differences in neighborhood retail services, we should begin by measuring and describing these differences. The current article moves the literature toward this end in two ways. First, we explore a number of approaches to measuring neighborhood “retail access” and compare various metrics over time and space. Second, whereas many previous studies focus on a single type of good or service, we consider access to a wide range of retail and household services. We combine publicly available data on business establishments at the ZIP code level with several New York City-specific data sources, including the location of all commercial and residential buildings, the location and corporate affiliation for several retail and food service chain establishments, residential population characteristics, and sales values of residential properties. Using this combined data set, we assess differences in retail access for approximately 208 neighborhoods, based on income and racial composition and differential growth rates in housing values.

Results confirm some findings from previous research, namely that lower income and minority neighborhoods have a lower density of commercial establishments and employment, smaller average establishments, a higher proportion of “unhealthy” restaurants, and, in some cases, less diversity across retail subsectors. However, the patterns vary by retail type and demographics: The disparities are smaller for grocery stores, pharmacies, and clothing than for food service or for retail as a whole. Perhaps surprisingly, predominantly Hispanic neighborhoods have more diverse retail and food services and greater access to retail corridors than predominantly White neighborhoods; the opposite is true for predominantly Black neighborhoods. Although most neighborhoods in New York City saw an increase in retail activity between 1998 and 2007, the rate of retail growth has been particularly fast in neighborhoods that were initially lower valued and experienced relatively high housing price appreciation compared with the city overall. However, initially higher valued and appreciating neighborhoods experienced relatively faster growth in the size of retail establishments.

This article proceeds in the following way. The next section provides a review of the relevant literature. The “Method and Data Description” section describes the data and methodology and the “Results” section summarizes the results from the current analysis. The “Conclusions and Policy Implications” section concludes and offers some implications for designing economic development policies.

Literature Review

Because the size of retail market areas will vary by product type, we begin by reviewing the types of retail that are likely to serve primarily neighborhood markets. We then outline several possible reasons why retail activity may differ by underlying neighborhood characteristics: variation in store operating or set-up costs, variation in purchasing power and preferences among local consumers (or residents), and institutional factors or public policies that influence commercial activity.

Market Area and Product Type

The urban economics literature provides several models of firm decision making and retail location that provide a theoretical framework for why the size of retail markets will vary by product type.

[Hotelling's \(1929\)](#) simple spatial model of firm location suggests that the density of stores depends on a variety of factors, including fixed costs of the store, buyer density, and travel costs, all of which may vary by neighborhood economic conditions. One implication is that there will be different market sizes and, thus, different densities of store networks, for establishments selling various products, which will translate into a hierarchy of retail networks. Retail store networks will be denser (i.e., more locally based) for stores that have low fixed costs and sell goods that are highly standardized and frequently consumed, so that consumers will not be willing to travel long distances to purchase them ([B. Berry, 1967](#)).¹ Following [DiPasquale and Wheaton \(1996\)](#), we focus on the lowest geographical level of store networks: neighborhood stores whose customers are drawn primarily from within the immediate vicinity; these establishments will most likely reflect the composition of neighborhood residents. The goods most likely to be sold at neighborhood stores include groceries, health and beauty products, and general household items, such as cleaning and household supplies. These items are typically sold at grocery stores, supermarkets, convenience stores, pharmacies, and general merchandise stores. In addition to retail, some service establishments primarily serve the immediate neighborhood, namely laundry services, coffee shops, and limited service restaurants, gyms, and beauty salons.

Variation in Store Fixed or Operating Costs

Set-up and operating costs for retailers include a number of factors, some of which vary by neighborhood and others that are specific to the firm and therefore somewhat idiosyncratic.² For instance, rents are likely to be higher in high-income (or high-wage) neighborhoods, whereas insurance and security costs increase with neighborhood crime rates. Although wages for similar positions (sales clerk or shelf stocker) may be relatively similar across neighborhoods within the same metropolitan area, there is some anecdotal evidence that employee turnover or training needs are higher in low-income neighborhoods ([International Council of Shopping Centers, 2004](#)), increasing average labor costs in those areas. Two other fixed costs that are likely to vary across neighborhoods are local land use regimes (zoning of commercial uses) and characteristics of the local building stock. Specifically,

restrictions against or incentives for retail occupancy can increase or reduce costs associated with initial setup. Similarly, the inherent nature of the building stock will determine the feasibility and costs associated with adapting the particular retail business to the existing commercial space. For example, grocery stores often require enough space and a robust enough infrastructure to support freezers, whereas restaurants require venting from stoves and ovens ([Barragan, 2010](#); [International Council of Shopping Centers, 2004](#)). Availability of suitable land parcels for development may be particularly important for large chains that have a preferred (often low-density) model for their stores (i.e., Big Box).

Variation in Consumer Characteristics

For any given type of store/product, the Hotelling model implies that the density of store networks will be increasing in density of buyers. The stylized model assumes that buyers are uniformly distributed and have homogeneous preferences. In reality, it is unlikely that all residents of a single neighborhood have the same demand function, either based on income/ability to pay or preferences, so estimating the density of actual rather than potential buyers within a given geographic area becomes more complicated. [Waldfoegel \(2008\)](#) demonstrates that there is considerable heterogeneity across consumer preferences for services such as restaurants and media, and that preferences are strongly correlated with observable population characteristics, such as educational attainment and race/ethnicity.

This conclusion mirrors the findings of a sizable body of literature in public health that explores the differences in the locational decisions of food establishments across neighborhoods. [Powell et al. \(2007\)](#), [Zenk \(2005\)](#), and [Alwitt and Donley \(1997\)](#) demonstrate that various retailers (namely banks and supermarkets) opt not to locate in poorer ZIP codes even after controlling for purchasing power—leading the authors to conclude that retail locational decisions may hinge on a host of factors in addition to an area's market potential. Interestingly, Alwitt and Donley found that fast food restaurants were least likely to discriminate across neighborhoods, whereas [Block, Scribner, and DeSalvo \(2004\)](#) and [Lewis et al. \(2005\)](#) found that fast-food restaurants were more likely to locate in poorer, predominately minority neighborhoods.

A few other empirical studies relate retail markets to local characteristics. [S. T. Berry and Waldfoegel \(2003\)](#) find that as market size increases, the range of product variety and quality widens, and the number of high-quality products grows. A recent study by [Chapple and Jacobus \(2009\)](#) of retail change in the San Francisco Bay area finds that retail revitalization is most strongly associated with gains for middle-income neighborhoods. [Zukin et al. \(2009\)](#) conduct case studies of two gentrifying neighborhoods in New York City and find a large increase in the number of independently owned (or local chain) establishments in those neighborhoods, compared with a small increase in large chain stores.

Institutional Factors and Public Policies

The models described thus far assume only market factors in the determination of local retail markets, but a variety of federal, state, and local public interventions have been used to try to stimulate business development and job growth, particularly in lower income urban neighborhoods.

The largest federal policy aimed at business development, the Empowerment Zones and Enterprise Communities Initiative, provides federal tax incentives and other financial benefits to businesses that locate in more than 100 designated neighborhoods in economically and socially depressed urban or rural communities ([Hebert, Vidal, Mills, James, & Gruenstein, 2001](#)). Empirical analysis of the Empowerment Zones/Enterprise Communities program provides very mixed evidence on the program's effectiveness, but most find little or no effect on job or firm creation ([Bondonio, 2003](#); [Busso & Kline, 2008](#); [Coopers & Lybrand Economic Studies Group, 1982](#); [Dowall, 1996](#); [Glickman, 1981](#); [Neumark & Kolko, 2008](#)).

The Community Development Block Grant program provides federal funds that can be spent on a variety of activities designed to enhance neighborhood economic and social conditions broadly, including economic development programs. As with the Empowerment Zones evaluations, evidence on the effects of Community Development Block Grant on employment growth in targeted neighborhoods is somewhat mixed: Higher spending per poor resident on economic development does seem to increase the number of businesses, but the effects vary by initial city and neighborhood conditions ([Galster, Walker, Hayes, Boxall, & Johnson, 2004](#); [Walker, Hayes, Galster, Boxall, & Johnson, 2002](#)).

Many local governments in large cities, including New York City, have additional policies designed to encourage business creation or retention in targeted areas, such as tax abatements. Local governments can also change the feasibility and costs of commercial activity indirectly through zoning codes by differentially allowing or restricting the uses and sizes of buildings.

Method and Data Description

In this article, we use data from the Census Bureau's ZIP Business Pattern series (an extension of the County Business Patterns data), as well as two New York City-specific data sets on commercial properties and chain establishments, to develop a set of metrics that describe neighborhood retail access. We then use those metrics to establish some stylized facts about the relationship between retail activity, income, and ethnic composition in New York City, and how retail activity changes over time in the context of neighborhood economic transition. The first part of the empirical analysis develops several different metrics, presenting summary statistics for each and examining the correlation between them, to determine whether the choice of metric is likely to affect the outcome of analysis. In the second part of the analysis, we present descriptive statistics around two research questions:

- 1.

How do retail patterns in New York City vary by neighborhood income and racial/ethnic composition?

Do these patterns differ by retail category?

2.

How has retail activity in New York City changed over time? How do the changes vary by baseline neighborhood economic characteristics and economic growth?

Development of Retail Metrics

Our primary source of data on retail and commercial activity is the ZIP Business Patterns (ZBP) data set, collected annually by the Census Bureau. The ZBP data provide counts of the number of establishments in each industrial sector, broken out in several size categories based on the number of employees.³ We are using the data from 1998 through 2007, which uses the North American Industrial Classification System (NAICS) to indicate industrial sector up to a 6-digit level of detail.⁴ Because our research focuses on retail that primarily serves the residents of the immediate neighborhood, and because we are interested in quality-of-life implications, we have chosen to focus on four industry categories that meet these criteria: supermarkets (NAICS 6-digit code: 445110), pharmacies and personal care stores (NAICS 3-digit code: 446), clothing stores (NAICS 3-digit code: 448), and food service establishments (NAICS 3-digit code: 722). To provide some context, we also look at the total number of establishments in retail (NAICS 2-digit code: 44-45) and food service and hospitality (NAICS 2-digit code: 72).

For each of these industrial groupings, we construct four metrics at the ZIP-code level. We match each ZIP code to the land area of the ZIP-code tabulation area (ZCTA) from the 2000 census, which allows us to calculate the density of establishments per acre, by industry-ZIP-year (because land areas vary widely by ZCTA, comparing simple counts of establishments across ZCTAs may be misleading).⁵

Second, we estimate the total employment by industry-ZIP-year, using the counts of establishments in each size category, and again use the land area to calculate employment density.⁶ Third, we combine the employment and establishment counts to calculate the average size of establishments by number of employees. Fourth, to measure the diversity of establishments, we construct a set of Herfindahl indices for each grouping. The Herfindahl index is calculated according to the following equation:

$$H = \sum_{i=1}^n s_i^2,$$

where s_i is the share of establishments in category i for a given industrial grouping. The index values range from 0 to 1, with higher numbers indicating greater concentration or less diversity. For example, if all the establishments in a ZIP code were in the same industry, then the share for that industry would be equal to 1, as would the value of the index. For the retail Herfindahl index, we use the share of establishments in each of the 12 three-digit NAICS categories within the two-digit retail industry (NAICS

code: 44-45). For the food service Herfindahl index, we use each of the 4 four-digit categories within the three-digit food service category (NAICS code: 722). The subcategories within each index are shown in [Table 1](#). All ZBP-based metrics are constructed as averages across all years in the sample (1998-2007) at the industry-ZIP level.⁷

Table 1. Retail and Food Service Subsectors

NAICS code	Industry subsector
44	Retail
441	Motor vehicle and parts dealers
442	Furniture and home furnishings stores
443	Electronics and appliance stores
444	Building material, garden equipment
445	Food and beverage stores
446	Health and personal care stores
447	Gasoline stations
448	Clothing and clothing accessories stores
451	Sporting goods, hobby, book, and music stores
452	General merchandise stores
453	Miscellaneous store retailers
454	Nonstore retailers
722	Food service
7221	Full-service restaurants
7222	Limited-service eating places

NAICS code	Industry subsector
7223	Special food services
7224	Drinking places (alcoholic beverages)

Source. North American Industrial Classification System (NAICS).

Our fifth metric of retail access draws on a different data source and focuses on geographic distance between residential and commercial properties. Using 2006 property-level data from the NYC Department of Finance Real Property Assessment Database and the NYC Department of City Planning PLUTO Database, which identify the location and use type of all properties in the city, we calculate the share of all residential building areas within $\frac{1}{4}$ mile and $\frac{1}{2}$ mile of a commercial corridor. Using geographic information system (GIS) mapping techniques, commercial corridors are identified as clusters of retail building area and properties classified under commercial zoning overlays (that permit retail use in mixed-use areas). A map of the retail corridors is displayed in [Figure A1](#) in the appendix. The ZBP has two main advantages as a data source: Because it is collected annually, it can be used to examine changes over time, and it is available for all ZIP codes across the country, allowing consistent analysis for multiple cities. A notable drawback to the ZBP, however, is that it provides no information on the type or quality of goods and services within each industrial category. For instance, one of the categories of interest is supermarkets. The 6-digit NAICS code for this category (445110) captures a wide range of store sizes and types, from branches of large national chains such as Safeway and Kroger, to small, independently-owned neighborhood stores or bodegas.⁸ Although these stores will overlap somewhat in goods offered, bodegas generally carry a much narrower range of products than traditional supermarkets, and may differ from supermarkets (and from one another) by quality and price. The same is true for our other categories of interest (notably food service).

To examine differences across neighborhoods in the quality of goods and services, we supplement the ZBP data with information on the location and corporate affiliation for a large number of regional and national chains in New York City, collected by the [Center for an Urban Future \(2009\)](#). From this database, we identify 98 chains that correspond to our categories of interest: all food services, some clothing and apparel (selected to cover a range of price points, adults of both genders and children, and with a large enough number of franchises to offer room for spatial variation), pharmacies, tax preparation services, gyms, and some home goods. The full list of chains selected is shown in [Table A1](#) in the appendix. We then aggregate the data to the ZIP-code level, calculating the total number of chain establishments and, within that, the number of chain restaurants. Of the restaurants, we flag certain chains as “unhealthy” fast food (shown in [Table A1](#) in the appendix), and calculate the share of chain restaurants that are “unhealthy” ([Neal, 2006](#); [Pillsbury, 2010](#); [Warde, Martens, & Olsen, 1999](#)). For

illustration, we also identify four iconic chains—McDonalds, Subway, Dunkin Donuts, and Starbucks. McDonalds is the most prevalent fast-food chain (it is third out of all chains, after Dunkin Donuts and Subway) with predominately unhealthy food choices. Subway also has a large number of locations and advertises itself as a healthier alternative that is still low cost. Dunkin Donuts and Starbucks both offer coffee and baked goods (and we make no claims about their comparative health values), but at different price points, in quite different environments, and their marketing strategies target different clientele. To identify some quality differences within our category of greatest interest, supermarkets, we augment data from the Center for an Urban Future chain database with online searches to assemble a list of locations for a large number of multi-establishment grocery store firms present in New York City. The list of firm names is shown in [Table A2](#) in the appendix. We use this to calculate the number of chain supermarkets in each ZIP code and identify several chains as “upscale”; these chains typically carry more organic or locally provided foods, have a large fresh-produce section, and offer hard-to-obtain or expensive specialty items.

One of the purposes of this article is to determine what types of metrics should be used to describe retail access and whether the choice of metric is likely to affect the patterns observed. Having created five separate metrics for several industry categories, we calculate pairwise correlations between all the retail metrics (shown in [Table 4](#) later in the text).

Testing the Relationship Between Retail Activity and Neighborhood Characteristics

To identify patterns of retail activity across neighborhoods in New York City, we calculate summary statistics for each of the metrics described above and compare them in several ways. As described in the “Literature Review” section, we expect that the amount and type of retail activity will vary by purchasing power, consumer preferences, and store costs. Thus, we will compare retail metrics across neighborhoods with underlying differences in variables ([Table 2](#)) that proxy for purchasing power and preferences, specifically household income and racial/ethnic composition.⁹ We also briefly address one potential difference in store costs; this part of the analysis is discussed in the next section. All data on population characteristics for ZIP codes are taken from the 2000 census of population and housing. To account for variation in size across the ZIP codes, we weight the summary statistics by population. As shown in [Table 3](#), ZIP-code areas in New York City exhibit significant differences from one another in underlying population characteristics, such as population density, income, and ethnic composition. The measures of retail activity and access also vary substantially across neighborhoods. In addition, the average population of a ZIP code in New York City is approximately 43,000, large enough to be a market area for neighborhood stores.

Variable	Definition	Source
Primary retail metrics		
Est/land	Number of establishments in ZIP divided by land area (acres)	ZBP (1998-2007), Census 2000
Emp/land	Number of establishments in ZIP divided by land area (acres)	ZBP (1998-2007), Census 2000
Emp/est	Total employment in ZIP divided by total establishments	ZBP (1998-2007)
Herfindahl	Herfindahl index of diversity for retail, food service	ZBP (1998-2007)
Share res within ¼mile commercial	Percentage of residential sq. ft. within ¼, ½ mile of commercial corridor	GIS calculations using RPAD
Chain retail metrics		
Chain stores	Number of retail establishments in selected chains	CUF (2009)
Chain restaurants	Number of restaurant establishments in selected chains	CUF (2009)
Percentage unhealthy	Number of fast-food chains/total chain restaurants	CUF (2009)
Gyms	Number of establishments in selected gym chains	CUF (2009)
Chain groceries	Number of establishments in selected grocery chains	Authors' search online
Percentage upscale	Number of upscale groceries/number of chain groceries	Authors' search online
Population characteristics		
Medium-upper income	ZIP income \geq 80% of NYC average household income	Census (2000)
Low income	ZIP income <80% of NYC average household income	Census (2000)

Variable	Definition	Source
Non-Hispanic White	>60% ZIP population non-Hispanic White	Census (2000)
Black	>60% ZIP population Black	Census (2000)
Hispanic	>60% ZIP population Hispanic (all races)	Census (2000)
Low value	ZIP average (residential) sales price < 80% NYC average price	DoF residential sales data (1998), Furman Center
Upgrading	ZIP percentage change in average housing value > NYC percentage change in average housing value (1998-2007)	DoF residential sales data (1998-2007), Furman Center
Stable/lagging	ZIP percentage change in average housing value \leq NYC percentage change in average housing value (1998-2007)	DoF residential sales data (1998-2007), Furman Center

Note. ZBP = ZIP business patterns; CUF, Center for an Urban Future; DoF = Department of Finance; GIS = geographic information system; RPAD = Real Property Assessment Database.

Table 3. Summary Statistics of All Variables

Variable	Mean	SD	Minimum	Maximum	<i>n</i>
Population	42,762	26,956	16	106,415	208
Pop/acre	62.35	44.70	0.13	277.81	208
Average household income	\$81,315	\$41,431	\$0	\$227,494	208
BA plus	36.6	26.9	0.0	100.0	208
Poverty rate	18.3	12.7	2.5	100.0	208
Percentage White	43.9	29.9	0.4	100.0	208
Percentage Black	21.8	26.6	0.0	93.9	208

Variable	Mean	SD	Minimum	Maximum	<i>n</i>
Percentage Hispanic	22.0	19.2	0.0	79.9	208
Sales price per unit, 1998	\$252,010	\$231,578	\$24,931	\$2,130,190	171
Percentage change in sales prices, 1998-2007	138.2	160.5	-26.6	1354.2	170
Est/land (retail)	0.26	0.46	0.00	2.79	208
Emp/land (retail)	2.69	5.94	0.00	48.36	208
Emp/est (retail)	9.78	8.28	2.50	95.25	208
Herfindahl (retail)	0.24	0.19	0.10	1.00	208
Share of residence within ¼ mile	0.85	0.22	0.02	1.00	163
Share of residence within ½ mile	0.93	0.17	0.10	1.00	124
Chain stores (per acre)	0.04	0.06	0.00	0.36	208
Chain restaurants (per acre)	0.02	0.05	0.00	0.29	208
Percentage unhealthy	0.23	0.16	0.00	0.75	208
Gyms (per acre)	0.00	0.00	0.00	0.04	208
Chain groceries (per acre)	0.00	0.01	0.00	0.04	208
Percentage upscale	0.03	0.11	0.00	0.80	208

Note. Population statistics are from 2000; business statistics are ZIP-industry averages across 10 years of ZIP business patterns (ZBP) data (1998-2007); chain statistics are calculated as of 2009.

To understand how differences in two key characteristics, income and racial/ethnic composition, affect patterns or retail activity, we compare retail metrics by these characteristics. Specifically, we compare the average value for each of our retail metrics (density of establishments and employment, average size, Herfindahl index, residential access, and the counts of various chains) for ZIPs in which average household income is less than 80% of the average income for New York City with ZIPs with average

household income above 80% of the city average income (\$58,505 in constant 2,000 dollars). To assess the correlation with ethnic composition, we compare retail metrics for ZIPs that are super-majority non-Hispanic White ($\geq 60\%$) with ZIPs that have a super majority of Black or Hispanic residents.

Changes in Retail Activity Over Time

In addition to comparing level differences in retail metrics, we are interested in how retail presence has changed over time and whether those changes reflect underlying changes in neighborhood characteristics. Thus, we calculate the growth rate for each of our retail metrics from 1998 to 2007 (annual changes tend to be quite small and somewhat noisy). The growth rate is calculated using a standard measure:

$$g_{i,98-07} = \frac{(\text{Retail}_{i,07} - \text{Retail}_{i,98})}{0.5 \times (\text{Retail}_{i,07} + \text{Retail}_{i,98})},$$

in which $\text{Retail}_{i,98}$ is the retail metric in industry i in 1998 and $\text{Retail}_{i,07}$ is the retail metric in industry i in 2007. As discussed in several previous articles that have used this measure, this growth rate provides a symmetric growth rate. By using an average of retail metrics in the beginning and ending years rather than just the beginning year in the denominator, we reduce potential measurement error associated with large deviations from average retail activity (see [Davis, Haltiwanger, & Schuh, 1996](#); [Haltiwanger, Jarmin, & Krizan, 2010](#), for more discussion). To analyze these changes in retail, we stratify the sample of ZIPs in two ways: by initial economic status as of 1998 and by transition in economic status between 1998 and 2007. Since we do not have any measures of income at the ZIP-code level after 2000, we use instead residential housing sales data to identify the initial and change in economic status for each neighborhood in New York City.¹⁰ We obtain NYC Department of Finance residential sales data for all New York City ZIP codes from the Furman Center for Real Estate and Urban Policy at New York University. The data provide us with the average price per unit for all residential sales transactions in New York City between 1998 and 2007. From these data, we construct relative measures of neighborhood housing values for every neighborhood i in year t :

$$\text{RelativeZIP_AvgPricePU}_{i,t} = \text{ZIP_AvgPricePU}_{i,t} / \text{NYC_AvgPricePU}_t.$$

To differentiate neighborhoods based on their initial economic status, we calculate this ratio for all ZIP codes in 1998 and classify neighborhoods with relative average housing values less than 0.8 as “Low

Value.” Neighborhoods with relative average housing values greater than or equal to 0.8 are classified as “Moderate/High Value.”

We then classify the neighborhoods based on their relative change in housing values between 1998 and 2007. We classify neighborhoods as “Upgrading” if they experience a percentage gain in average housing values (absolute, not relative) that is greater than the percentage change in average housing values for the city overall (housing values for New York City on average increased by 120% between 1998 and 2007). Neighborhoods with percentage changes in average housing values less than those experienced for the city overall are classified as “Stable/Lagging.”

Since we are interested in observing how changes in retail manifest themselves in low-income neighborhoods, we first compare changes in the retail metrics across low-value neighborhoods that are upgrading and stable/declining. In addition, we want to compare changes in retail activity with those experienced by relatively higher income neighborhoods. Therefore, we also calculate the difference in change between low-value upgrading and stable/declining neighborhoods and compare this with the same difference across moderate-/high-value neighborhoods. The initial retail landscapes are quite different across low- and moderate-/high-value neighborhoods (retail activity is significantly lower in low-value neighborhoods), and therefore we conduct this simplified “difference-in-difference” to avoid any upward bias in our estimates of retail change in low-value neighborhoods.

Additional Data Issues

A possible concern with the ZBP data is the consistency of the industrial classification system. According to the census, in the surveys used to construct the ZBP database, establishments are self-classified by employee or contact at the company, based on revenues. This raises the possibility that similar types of establishments may be classified differently, particularly for establishments engaged in multiple activities. For instance, as described above, bodegas could be classified as grocery stores because they sell food items. Many, but not all of these stores, also prepare and sell some fresh food, such as made-to-order deli sandwiches, coffee, and bagels. Depending on the share of revenues received from these activities (or the knowledge of the employee filling out the survey), a bodega may be counted in the ZBP as either a grocery store (NAICS code: 4451) or a limited service restaurant (NAICS code: 7222), while performing largely similar functions. Similar ambiguity in the NAICS codes may be present for other types of establishments as well. Classification of the same establishment may also change over time, even if the establishment does not change functions.¹¹

Results

In this section, we provide summary statistics of the various retail metrics to describe the amount, type, and mix of retail access and activity in New York City. We describe variation in retail metrics

across neighborhoods by income and ethnic composition, and changes over time, stratified by economic growth.¹²

How Can We Characterize Retail Activity in New York City?

All neighborhoods in New York City have at least some amount of retail activity; there are no entirely residential neighborhoods in the city. However, there is considerable variation in the quantity of retail activity, measured by the density of establishments and shown in [Figure 1](#). [Table 3](#) also displays a selection of the retail metrics and demonstrates that neighborhoods across the city are diverse in terms of retail density, the size of retail establishments, and the diversity of stores and services. Although residents, on average, have great access to retail (nearly 90% of all residential space is within ¼ mile of a retail corridor), there are neighborhoods where this is the case for less than 10% of residentially occupied space.

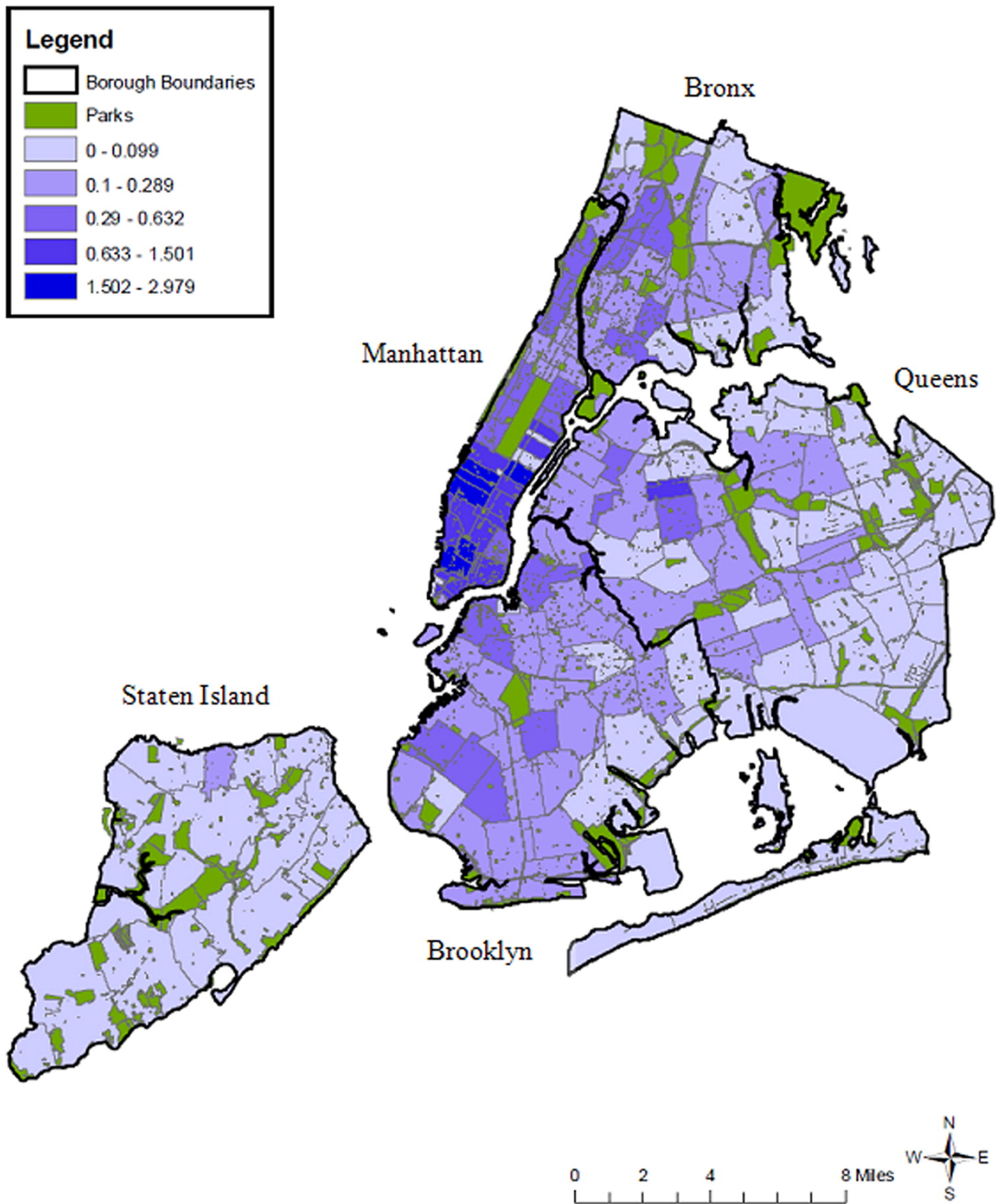


Figure 1. Density of retail establishments by ZIP code (2007)

Because one of the purposes of this analysis is to determine what metric or combination of metrics should be used to characterize retail, it is worth asking to what degree the various metrics are correlated with one another, both within and across industry categories. If the retail metrics are not strongly correlated, that implies that the choice of metric (in our analysis, intended to serve as the

dependent variable) may affect the results of the analysis. In [Table 4](#), we show simple pairwise correlation coefficients between each metric for groceries, all retail establishments, all food service establishments, and chains. In general, the measures of establishment density across the categories are highly correlated (ranging from 0.70 to 0.90), which suggests that there is colocation among different types of retail. However, the other metrics, such as size, are not as consistently or strongly associated. Moreover, neither retail diversity (as measured by the Herfindahl index) nor retail access (as measured by distance to a retail corridor) is highly correlated with the other metrics. This suggests the need for a multi-dimensional approach to characterizing neighborhood retail activity.

Table 4. Correlation of Selected Retail Metrics

Variable	Grocery			All retail			Food service	
	Est/land	Emp/land	Emp/est	Est/land	Emp/land	Emp/est	Est/land	Emp/land
Grocery								
Emp/land	0.592							
Emp/est	-0.150	0.461						
All retail								
Est/land	0.729	0.576	0.015					
Emp/land	0.615	0.643	0.133	0.935				
Emp/est	-0.079	0.333	0.540	-0.007	0.082			
Food service								
Est/land	0.698	0.663	0.087	0.901	0.870	0.033		
Emp/land	0.569	0.630	0.142	0.822	0.864	0.079	0.917	
Emp/est	0.191	0.437	0.284	0.174	0.250	0.667	0.214	0.321

Variable	Grocery			All retail			Food service	
	Est/land	Emp/land	Emp/est	Est/land	Emp/land	Emp/est	Est/land	Emp/land
Herfindahl retail	0.259	0.261	0.100	-0.064	-0.023	0.097	0.000	0.033
Chain store density	0.660	0.582	0.044	0.824	0.841	0.028	0.905	0.853
Percentage unhealthy	-0.136	-0.275	-0.188	-0.203	-0.236	-0.188	-0.256	-0.241
Starbucks-Dunkin Donuts	0.484	0.654	0.263	0.666	0.702	0.246	0.747	0.823
Percentage residence units within ¼ mile of retail corridor	0.477	0.253	-0.269	0.312	0.244	-0.218	0.282	0.211

Note. Correlation coefficients for ZIP Business Patterns (ZBP) metrics are from 2007.

How Does Retail Activity in New York City Vary by Income?

To develop a better understanding of the relationships between household income and retail activity, we compare all the retail metrics for neighborhoods with average household income above and below 80% of the city average income in 2000 (\$58,505).¹³ In [Table 5](#), we display the results for the primary retail and food service metrics, which summarize overall retail access by neighborhood type.

Consistent with theory and previous case studies, relatively higher income neighborhoods have higher densities of both establishments and employment and larger establishments on average (for retail and food service). These disparities are generally larger for food services. In addition, the Herfindahl index for food services is significantly lower, on average, in higher income neighborhoods, suggesting that they have access to a more diverse pool of food services (the Herfindahl index for retail services is also

lower in higher income neighborhoods, but this difference is statistically insignificant). Although both low- and moderate-/high-income neighborhoods have considerable access to retail corridors, low-income neighborhoods have significantly more access. Together, these results suggest that residents in relatively low-income neighborhoods have retail activity nearby, but that it is less dense and composed of smaller and less diverse options (both of which could have implications for the quality and cost of the goods and services).

Table 5. Retail Access: Primary Retail Metrics by Income

	NYC	Middle-upper income	Low income	Difference
Est/land				
Retail	0.256	0.269	0.216	0.053 ^{***}
Food service	0.136	0.153	0.082	0.070 ^{***}
Emp/land				
Retail	2.488	2.821	1.421	1.400 ^{***}
Food service	2.166	2.641	0.650	1.991 ^{***}
Emp/est				
Retail	9.208	9.897	7.032	2.865 ^{***}
Food service	11.582	12.574	8.509	4.066 ^{***}
Herfindahl				
Retail	0.232	0.231	0.237	-0.006
Food service	0.448	0.444	0.460	-0.016 ^{***}
Share of residence within				
¼ mile	0.824	0.854	0.933	-0.079 ^{***}
½ mile	0.948	0.944	0.967	-0.024 ^{***}

	NYC	Middle-upper income	Low income	Difference
<i>n</i>	208	169	39	

Note. “Middle-upper income” defined as greater than 80% of NYC average household income. Statistics are population-weighted ZIP-industry averages across 10 years of ZBPA data (1998-2007).

*
 $p < .10$. ** $p < .05$. *** $p < .01$.

In the “Literature Review” section, we discussed how neighborhood characteristics other than income, such as access to transit and the amount of retail space in the local building stock, may affect variation in retail density. To test the importance of these factors, we calculate the average proximity between subway and rail transit and commercial properties and the average retail space per building for each ZIP.¹⁴ Although [Table 5](#) does not display these statistics, we find that relatively low-income neighborhoods have greater access to transit and more retail space per building. Therefore, in spite of possessing some characteristics that would, theoretically, make these neighborhoods more appealing to retail businesses, they still face less retail access overall.

Next we drill down to finer industry categories to better understand variations in access to specific quality-of-life retail. Again, we compare retail activity across low- and middle/high-income neighborhoods. These results are displayed in [Table 6](#). Overall, the pattern echoes that for retail access more generally. The density (in terms of establishments and employment) is significantly higher in relatively high-income neighborhoods, and the magnitude is the largest for food services; none of the other differences compare, in terms of magnitude, with the differences found among retail establishments more broadly (which are consistently much larger). The retail establishments are also larger in higher income neighborhoods, a pattern that is particularly stark for grocery stores. The grocery stores, however, are slightly denser in lower income neighborhoods, and there is no statistically significant difference in the density of large grocery stores across the two types of neighborhoods (this result, however, is not replicated using the employment density measure). Therefore, it appears that lower income neighborhoods are not as severely disadvantaged when it comes to “necessity” services and goods, such as groceries and drugstores; restaurants and food establishments, however, disproportionately locate in higher income neighborhoods.

Table 6. Quality-of-Life Retail Metrics by Income

	NYC	Middle-upper income	Low income	Difference
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	NYC	Middle-upper income	Low income	Difference
Est/land				
Groceries	0.040	0.036	0.051	-0.015***
Large groceries	0.004	0.004	0.002	0.002***
Drugstores	0.021	0.022	0.018	0.004***
Clothing	0.039	0.043	0.027	0.016***
Food service	0.136	0.153	0.082	0.070***
Emp/land				
Groceries	0.441	0.475	0.343	0.132***
Large groceries	0.291	0.339	0.153	0.186***
Drugstores	0.272	0.313	0.153	0.161***
Clothing	0.531	0.629	0.238	0.390***
Food service	2.166	2.641	0.650	1.991***
Emp/est				
Groceries	12.76	14.59	7.53	7.07***
Drugstores	76.61	80.69	63.91	16.78***
Clothing	12.99	14.57	8.47	6.10***
Food service	10.77	11.50	8.60	2.90***
<i>n</i>	208	169	39	

Note. "Middle-upper income" defined as greater than 80% of NYC average household income. Statistics are population-weighted ZIP-industry averages across 10 years of ZBPA data (1998-2007).

*
 $p < .10$. ** $p < .05$. *** $p < .01$. (Based on *t* test for difference in means.)

Finally, we compare the prevalence of chain stores and restaurants across low- and middle-/high-income neighborhoods. The results in [Table 7](#) indicate that higher income neighborhoods have more chain stores and restaurants. In addition, chains in poorer neighborhoods tend to be more unhealthy (35% compared with 21% of chain restaurants in higher income neighborhoods). Note that the difference in McDonalds' locations is about two times that for Subway locations. Dunkin Donuts are more prevalent than Starbucks in both types of neighborhoods, but the Starbucks–Dunkin Donuts ratio is the highest in middle-/high-income neighborhoods (0.54 compared with 0.03 in low-income neighborhoods and 0.41 citywide). Although we cannot make a clear-cut comparison on the health aspects of Starbucks versus Dunkin Donuts, the latter certainly offers a lower cost option and markets itself to a different clientele. The difference in the number of gyms is also stark: Middle-/high-income neighborhoods have, on average, one gym, whereas low-income neighborhoods have just 0.29. This result, together with the statistics on chain food stores, conveys a rather unhealthy environment for residents in low-income neighborhoods. Finally, we see that lower income neighborhoods actually have significantly more chain supermarkets, but they have no “upscale” markets. Although poorer neighborhoods likely do not suffer from lack of access to a Whole Foods or Zabar’s, this may point to disparities in access to fresh produce or other healthy foods.

Table 7. Chain Stores and Restaurants by Income

	NYC	Middle-upper income	Low income	Difference
Chain stores (<i>n</i>)	22.87	24.11	18.90	5.21 ^{***}
Chain restaurants (<i>n</i>)	15.27	16.01	12.90	3.11 ^{***}
“Unhealthy” (%)	24.7	21.1	35.2	−14.2 ^{***}
Notable chains (percentage of total)				
McDonalds	6.8	5.8	9.7	−0.04 ^{***}
Subway	8.4	8.0	9.8	−0.02 ^{***}
Starbucks	3.8	5.0	0.4	0.05 ^{***}
Dunkin Donuts	11.8	12.2	10.7	0.01 ^{***}
Starbucks–Dunkin Donuts ratio	0.41	0.54	0.03	0.50 ^{***}

	NYC	Middle-upper income	Low income	Difference
Gyms (<i>n</i>)	0.86	1.04	0.29	0.75***
Chain groceries (<i>n</i>)	3.34	2.92	4.67	-1.75***
“Upscale” (%)	3.2	4.4	0.0	4.4***
<i>n</i>	208	169	39	

Note. “Middle-upper income” defined as greater than 80% of NYC average household income. Statistics are population-weighted ZIP-industry averages across 10 years of ZBPA data (1998-2007).

* $p < .10$. ** $p < .05$. *** $p < .01$. (Based on *t* test for difference in means.)

How Does Retail Activity in New York City Vary by Race/Ethnic Composition?

We replicate the same analyses across neighborhoods stratified by supermajority race/ethnicity. [Table 8](#) displays the results for the primary retail access metrics. Consistent with the income results, neighborhoods with predominantly Black and Latino residents have lower establishment and employment densities and smaller establishments than those with predominantly White residents. Predominantly White neighborhoods also have significantly more diverse food service activity than predominantly Black neighborhoods (as indicated by the lower Herfindahl index); retail diversity in Latino neighborhoods, however, is greater than that in predominantly White neighborhoods. In addition, residents in predominantly Hispanic neighborhoods on average live closer to retail corridors (the opposite is true for predominantly Black neighborhoods, compared with predominantly White neighborhoods). In sum, neighborhoods with predominantly minority populations (which also tend to be poorer) have access to fewer and smaller stores and services. Retail diversity and physical access to retail services, however, vary by race/ethnicity. Residents in predominantly Latino neighborhoods on average live closer to retail and have more diverse retail options; residents in predominantly Black neighborhoods, however, have less physical access to retail services and somewhat less diverse options.

Table 8. Retail Access: Primary Retail Metrics by Predominant Racial/Ethnic Group

	White non-Hispanic	Black	Hispanic	Difference, White-Black	Difference, White-Hispanic
Est/land					
Retail	0.336	0.129	0.263	0.207***	0.073***
Food service	0.210	0.037	0.076	0.173***	0.134***
Emp/land					
Retail	4.084	0.874	1.653	3.209***	2.431***
Food service	4.479	0.366	0.676	4.113***	3.804***
Emp/est					
Retail	10.570	7.195	6.478	3.375***	4.092***
Food service	14.963	11.423	9.111	3.540***	5.852***
Herfindahl					
Retail	0.265	0.245	0.219	0.020	0.046***
Food service	0.470	0.519	0.430	-0.049***	0.040***
Share of residence within					
¼ mile	0.873	0.851	0.960	0.023***	-0.086***
½ mile	0.947	0.942	1.000	0.005***	-0.053***
<i>n</i>	78	26	16		

Note. Predominant racial/ethnic group defined as greater than 60% of population. Statistics are population-weighted ZIP-industry averages across 10 years of ZBPA data (1998-2007).

*
 $p < .10$. ** $p < .05$. *** $p < .01$. (Based on *t* test for difference in means.)

As for access to transit and prevalence of retail space, predominantly Black neighborhoods generally have more rail access than and about the same amount of retail space per building as predominantly White neighborhoods. Predominantly Hispanic neighborhoods, however, have significantly less access to transit and more retail space per building. Again, it appears that any disadvantage (or advantage in the case of predominantly Hispanic neighborhoods) in retail access is not solely driven by these cost-related factors.

Table 9 shows the results for finer, quality-of-life retail categories and indicates that overall minority neighborhoods have relatively lower densities (for establishments and employment) and smaller establishments for local basic services such as groceries, drugstores, clothing, and food services. One exception is groceries, which are more densely located in Latino neighborhoods relative to predominantly White neighborhoods; they do tend to be smaller, which could represent mostly bodega or deli-type outfits rather than general supermarkets (on average 6 employees compared with about 19 in predominantly White neighborhoods). As with the income results, these differences tend to be the largest for food service establishments.

Table 9. Quality-of-Life Retail Metrics by Predominant Racial/Ethnic Group

	White non-Hispanic	Black	Hispanic	Difference, White-Black	Difference, White-Hispanic
Est/land					
Groceries	0.040	0.029	0.064	0.012 ^{***}	-0.023 ^{***}
Large groceries	0.007	0.002	0.003	0.005 ^{***}	0.004 ^{***}
Drugstores	0.030	0.012	0.022	0.018 ^{***}	0.007 ^{***}
Clothing	0.064	0.017	0.036	0.047 ^{***}	0.029 ^{***}
Food service	0.210	0.037	0.076	0.173 ^{***}	0.134 ^{***}
Emp/land					
Groceries	0.724	0.220	0.377	0.504 ^{***}	0.347 ^{***}

	White non-Hispanic	Black	Hispanic	Difference, White-Black	Difference, White-Hispanic
Large groceries	0.569	0.107	0.162	0.462 ^{***}	0.407 ^{***}
Drugstores	0.469	0.118	0.199	0.351 ^{***}	0.270 ^{***}
Clothing	1.067	0.134	0.298	0.933 ^{***}	0.770 ^{***}
Food service	4.479	0.366	0.676	4.113 ^{***}	3.804 ^{***}
Emp/est					
Groceries	19.19	8.86	6.25	10.330 ^{***}	12.94 ^{***}
Drugstores	91.58	72.99	61.20	18.586 ^{***}	30.38 ^{***}
Clothing	15.88	11.66	8.86	4.215 ^{***}	7.01 ^{***}
Food service	13.24	8.37	7.82	4.864 ^{***}	5.42 ^{***}
<i>n</i>	78	26	16		

Note. Predominant racial/ethnic group defined as greater than 60% of population. Statistics are population-weighted ZIP-industry averages across 10 years of ZBPA data (1998-2007).

*
 $p < .10$. ** $p < .05$. *** $p < .01$. (Based on *t* test for difference in means.)

The last set of results for chain stores and restaurants are displayed in [Table 10](#). Predominantly Black and Hispanic neighborhoods have significantly fewer chains and considerably more “unhealthy” chain restaurants (41% and 30% for Black and Latino neighborhoods, respectively, compared with 13% for predominantly White neighborhoods). Consistent with the income results, predominantly minority neighborhoods have fewer Starbucks relative to Dunkin Donuts, fewer gyms, and no “upscale” supermarkets. Predominantly Hispanic neighborhoods, however, have more chain groceries overall, although the previous set of results suggests that they are, on average, smaller.

Table 10. Chain Stores and Restaurants by Race/Ethnicity

	White non-Hispanic	Black	Hispanic	Difference, White-Black	Difference, White-Hispanic
Chain stores (<i>n</i>)	23.98	19.67	19.27	4.31 ^{***}	4.72 ^{***}
Chain restaurants (<i>n</i>)	15.71	13.11	12.44	2.60 ^{***}	3.27 ^{***}
“Unhealthy” (%)	12.8	40.6	29.6	-27.8 ^{***}	-16.8 ^{***}
Notable chains (%)					
McDonalds	4.2	7.7	9.4	-3.4 ^{***}	-5.1 ^{***}
Subway	7.8	7.1	11.3	0.7	-3.6 ^{***}
Starbucks	8.9	0.4	0.7	8.5 ^{***}	8.2 ^{***}
Dunkin Donuts	11.5	10.1	12.1	1.4 ^{***}	-0.6
Starbucks-Dunkin Donuts ratio	0.94	0.03	0.06	0.92 ^{***}	0.88 ^{***}
Gyms (<i>n</i>)	1.41	0.52	0.44	0.90 ^{***}	0.98 ^{***}
Chain groceries (<i>n</i>)	2.26	4.25	3.90	-1.99 ^{***}	-1.64 ^{***}
“Upscale” (%)	8.8	0.0	0.0	8.8 ^{***}	8.8 ^{***}
<i>n</i>	78	26	16		

Note. Predominant racial/ethnic group defined as greater than 60% of population. Statistics are population-weighted ZIP-industry averages across 10 years of ZBPA data (1998-2007).

*
 $p < .10$. ** $p < .05$. *** $p < .01$. (Based on *t* test for difference in means.)

How Does Retail Activity in New York City Vary Over Time?

In this section, we review the results from the dynamic analysis, which looks at changes in retail activity between 1998 and 2007 for low- and moderate-/high-valued neighborhoods undergoing economic transitions.¹⁵ The first column of [Table 11](#) shows the changes in four retail metrics for the city as a

whole, and columns 2 and 3 show the metrics for low-valued neighborhoods that are either upgrading or stable/lagging. The fourth and fifth columns display the difference in retail change across upgrading and stable/lagging neighborhoods, for both low-valued and high-valued neighborhoods. We present the statistics for four retail categories to pick up any variation among types of service.¹⁶

Table 11. Does Retail Access Improve in Low-Value Neighborhoods That Upgrade?

	Percentage change in housing values (1998-2007)			Difference, upgrading-stable/lagging	
	NYC	Upgrading	Stable/lagging	Low-value ZIPs	Moderate-/High- Value ZIPs
Percentage change est/acre					
Retail (44)	10.2	20.03	15.01	5.02	1.10
Food service (72)	29.3	44.31	32.53	11.78*	12.34**
Groceries	16.4	23.23	24.18	-0.94	-3.54**
Clothing	7.5	31.49	7.34	24.15**	22.36*
Percentage change emp/acre					
Retail (44)	19.4	31.95	18.08	13.87*	25.92***
Food service (72)	32.9	49.31	30.98	18.34*	16.20**
Groceries	15.9	30.09	6.01	24.08**	31.25**
Clothing	20.4	37.25	8.86	28.39	32.80*
Percentage change emp/est					
Retail (44)	9.8	12.95	4.05	8.90	25.27***
Food service (72)	5.3	7.62	1.14	6.48	4.24

	Percentage change in housing values (1998-2007)			Difference, upgrading-stable/lagging	
	NYC	Upgrading	Stable/lagging	Low-value ZIPs	Moderate-/High-Value ZIPs
Groceries	0.3	9.16	-17.40	26.56**	34.73***
Clothing	14.5	12.30	2.03	10.27	13.44
Percentage change Herfindahl					
Retail (44)	65.8	47.61	61.77	-14.16**	-6.83
Food service (72)	0.5	0.14	-0.39	0.53	-7.99**
<i>n</i>	171	31	64	95	76

Note. ZIPs in columns 1 and 2 had initial housing values (1998) <80% NYC average. “Upgrading” defined as ZIP percentage change in average housing value >NYC percentage change in average housing value (1998-2007). “Stable/lagging” defined as ZIP percent change in average housing value ≤NYC percentage change in average housing value (1998-2007).

*
 $p < .10$. ** $p < .05$. *** $p < .01$. (Based on *t* test for difference in means.)

Both upgrading and stable/lagging low-valued neighborhoods are growing in terms of retail activity. This is consistent with most of New York City, and the outer boroughs in particular, where most of the low-valued neighborhoods are located (see [Figure 2](#)). Moreover, upgrading neighborhoods are generally outpacing the stable/lagging neighborhoods and about half of these differences are statistically significantly different from zero. Upgrading neighborhoods are receiving significantly more food service and clothing establishments (per acre) compared with stable/lagging neighborhoods, but the change in density and size of retail establishments more generally is indistinguishable across the two types of neighborhoods (with the exception of employee density, which is marginally statistically significant). In addition, all low-valued neighborhoods are becoming more homogeneous in terms of retail services (the Herfindahl index is increasing), but the upgrading neighborhoods are doing so at a slower rate (there is no discernible difference in food service diversity). This reflects the general trend toward larger chains (that might actually offer a wider range of product options within each store location). That said, the more substantial changes in retail activity for economically upgrading

neighborhoods are consistent with increasing buying power and, perhaps, shifts in consumer preferences, that is, toward larger chain stores.

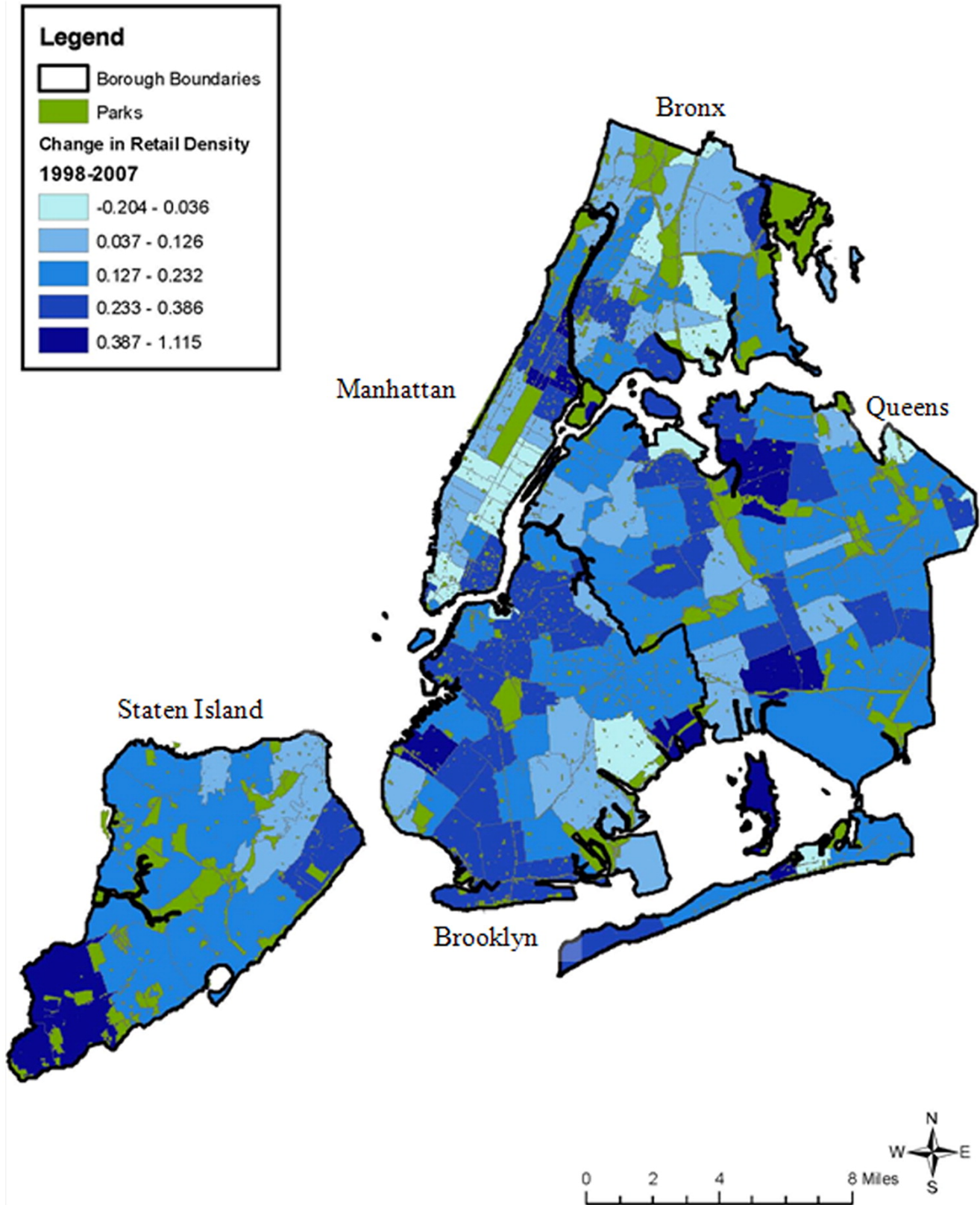


Figure 2. Percentage change in retail establishment density (1998-2007)

As a second comparison, we calculate the same statistics for moderate-/high-valued neighborhoods and derive the difference in retail change between the neighborhoods that were upgrading and stable/lagging (displayed in the last column of [Table 11](#)). We then compare this difference with the difference in retail change calculated for the low-valued neighborhoods (i.e., compare the two right-hand columns of [Table 11](#)). Overall, both low- and high-/moderate-valued neighborhoods exhibit the same change patterns across upgrading and stable/declining neighborhoods. However, in terms of establishment densities, low-valued and upgrading neighborhoods are generally outpacing their stable/lagging comparison neighborhoods more so than high-valued and upgrading neighborhoods (except for food services, which is growing at a relatively slower pace for low-valued and upgrading neighborhoods). The results for employee densities show the reverse: High-valued and upgrading neighborhoods are generally outpacing their stable/lagging comparison neighborhoods more so than low-valued and upgrading neighborhoods (again, with the exception of food services). Similarly, except for food service establishments, higher valued and upgrading neighborhoods are outpacing comparable stable/lagging neighborhoods in establishment size more so than low-valued neighborhoods. These results suggest that although lower valued neighborhoods are growing relatively faster in terms of retail establishment density, they are not attracting as many larger businesses (again, this might be due to differential location choices of larger chains). Last, although retail activity in high-valued and upgrading neighborhoods is becoming more homogeneous, it is doing so at a relatively faster pace than low-valued and upgrading neighborhoods (compared with similar stable/lagging neighborhoods). The opposite is true for diversity of food services. In sum, low-valued neighborhoods appear to fare better (in terms of retail activity) if they are economically upgrading, compared with similar, economically stable/lagging neighborhoods. However, initially higher valued and appreciating neighborhoods experienced relatively faster growth in the size of retail establishments.

Conclusions and Policy Implications

Scholars have dedicated a great deal of rigor and thought to understanding the nature of and mechanisms behind residential neighborhood change. Although we expect commercial amenities to change along with the residential population, and anecdotal evidence generally supports this, essentially no quantitative research has focused on neighborhood commercial change. In the current article, we aim to fill this gap by providing a much needed assessment of local retail establishments and the neighborhoods they serve. First, we construct and compare various metrics of “retail presence,” and find that a single measure is not sufficient for capturing the multidimensional nature of retail presence. Whereas the densities of establishments and employment are strongly and positively correlated, measures of size, access, and diversity are generally negatively and weakly correlated with density measures.

Second, we analyze how retail services vary across neighborhoods with different economic and demographic characteristics. Results show that low-income neighborhoods have lower densities of both establishments and employment, smaller average establishment size, and less diverse retail composition. However, the size of disparities varies by retail category: Poor neighborhoods are more disadvantaged in food service than in retail, and within retail, the differences are smallest for basic necessities, such as grocery stores and pharmacies. Low-income neighborhoods have fewer chain stores and restaurants, somewhat contrary to conventional wisdom. Supporting prior findings, a much higher proportion of chain restaurants in poor neighborhoods are unhealthy fast food establishments, and there are many fewer gyms available. Low-income neighborhoods actually have a higher number of chain supermarket branches, but are less likely to have upscale supermarkets, possibly a proxy for food quality. Similar disparities exist when comparing predominantly White neighborhoods with predominantly Black and Latino ones. Predominantly Latino neighborhoods, however, do have more diverse retail and food services and greater physical access to retail corridors than predominantly White neighborhoods.

Third, we examine changes in retail activity over time. We find that by almost all measures, retail access has improved in New York City between 1998 and 2007, and that it improved particularly rapidly in low-value neighborhoods that experienced upgrading or gentrification. However, initially higher valued and appreciating neighborhoods experienced relatively faster growth in larger retail establishments over this same time period. Together, these results suggest that retail is quite sensitive to changes in neighborhood economic or demographic characteristics.

A number of areas remain for future research. This article focuses exclusively on New York City, which differs from many other U.S. cities in its size, density, and integration of residential and commercial activities throughout many neighborhoods. Thus, an important next step is to conduct similar analyses and verify whether the relationships between income, race, and retail access hold true in other cities.¹⁷

Second, as new neighborhood-level data become available from the 2010 census (and the 2005-2009 average of the American Community Survey), it will become possible to examine more directly how retail patterns have changed in neighborhoods undergoing economic and demographic change. This may help illuminate some of the causes behind the disparities: If low incomes are the main source of limited retail access, then retail should increase in neighborhoods experiencing rising incomes.

Additional research in this area should help policy makers better construct programs to help residents in low-income and minority neighborhoods gain access to a broader range of goods and services, and healthier food options, and to generally improve neighborhood quality of life.

The purpose of this article is not to evaluate the success of existing economic development policies, but based on our results, we pose several questions about how future programs might be designed.

First, should eligibility be defined through constant geographic boundaries, as is the case for programs such as federal empowerment zones? As shown in [Figure 3](#), although there is some geographic clustering of poor and minority neighborhoods, rigid geographic boundaries are somewhat of a blunt

instrument if the intent is to target investment toward these neighborhoods. In addition, retail markets may span broader or more distinct areas than those defined by racial or economic clusters.

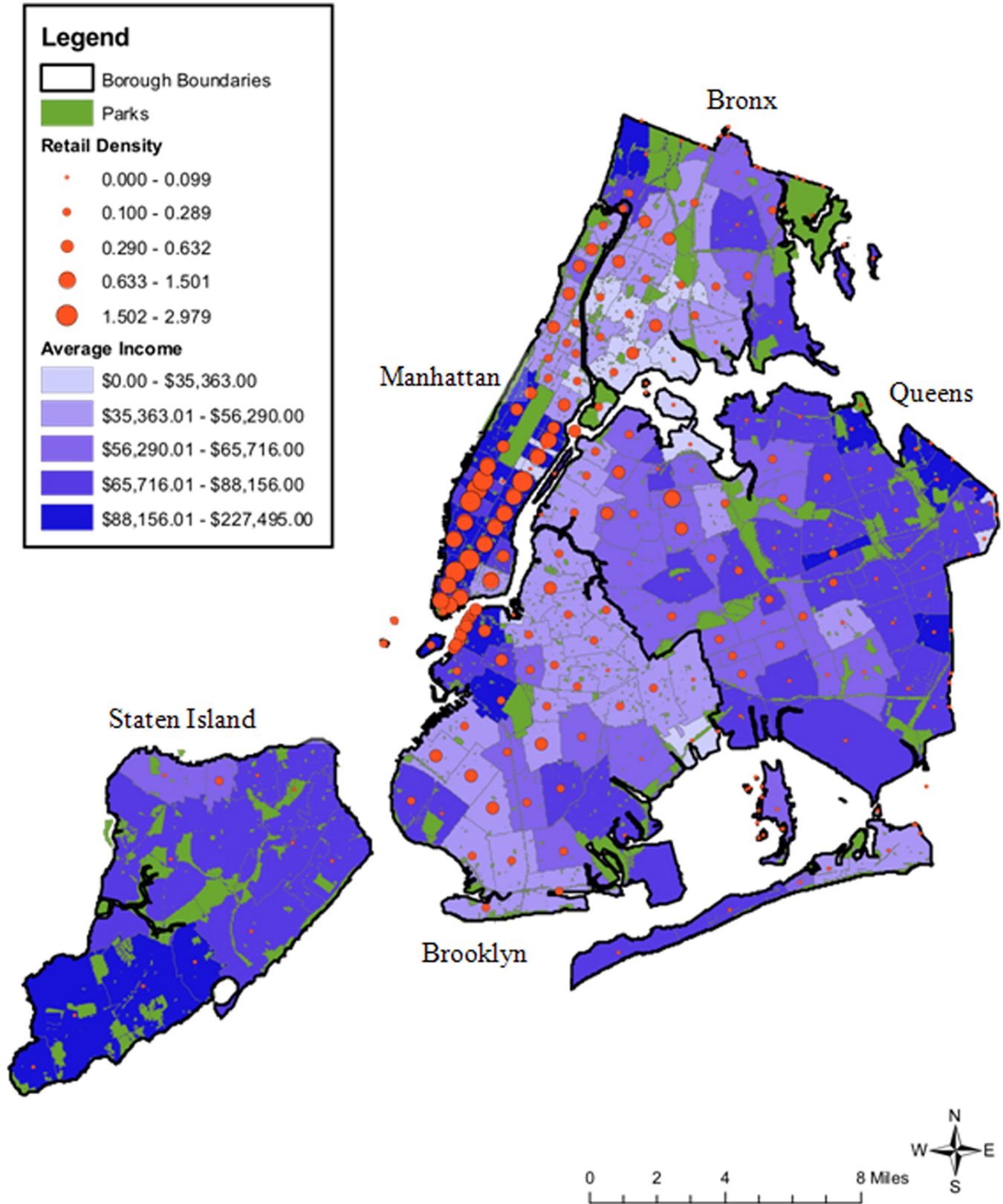


Figure 3. Density of retail establishments (2007) and average household income (2000)

A second question is whether policies should target or favor businesses based on size. Many traditional economic development policies are intended to help small, locally owned “mom and pop” businesses. Such assistance is thought to improve opportunities for entrepreneurs and assist wealth building within the community. However, our research finds that poor and minority neighborhoods currently have much smaller average stores for nearly all the retail categories examined. This implies a smaller range of product choices within each store, and, to the extent that economies of scale exist, may result in higher prices than in larger establishments. Moreover, larger stores by definition offer more opportunities for employment. Thus, although small business–friendly policies may be popular among business owners, it is not clear if they are advantageous for either consumers or potential workers in low-income neighborhoods. Alternatively, perhaps the public programs could subsidize certain types of products (e.g., healthier ones) for smaller businesses that cannot offer those goods at competitive prices.

Third, should economic development policies treat all types of commercial activity as equally desirable? Policies that encourage manufacturing or business incubation may provide employment opportunities but will not directly address the discrepancies in access to grocery stores, clothing, or healthy food service options. Targeted policies around health-related retail may be more effective at alleviating consumption disparities than more broadly framed tax abatements. Policy makers should consider whether the primary purpose of each program is to encourage jobs or benefit consumers, and whether these goals are mutually exclusive.

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Competing Interests

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Footnotes

1. In the classic example, the market area for ice cream vendors will be very small, due to the highly perishable nature of the good, so in equilibrium there will be a large number of vendors each with a small market area. On the other end of the spectrum, consumers should be willing to travel long distances to purchase goods that are expensive, infrequently purchased, or highly differentiated by quality, such as cars, furniture, or high-end restaurants; these retail categories will have a smaller number of establishments, each serving quite large geographic markets.

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2. Many retail firm costs are not only not “fixed” in the traditional sense but are also not exactly marginal. For instance, building rents are often fixed over lease terms, which may be 5 or 10 years long but may offer some flexibility between leases, depending on negotiations between tenant and landlord. Likewise contracts with suppliers, insurance, utilities, and so on, may be fixed over a short period of time (1-2 years), and so cannot be directly reduced with marginal productivity.

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3. An establishment is defined as a “single physical location at which business is conducted or services or industrial operations are performed.” A firm may have multiple establishments, each of which are counted separately.

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4. Prior to 1997, the ZBP use the Standard Industrial Classification system, which differs from the NAICS in several ways. In general, the NAICS offers a more fine-grained level of detail that is helpful for our analysis, but makes it infeasible to match counts by industry category precisely between the two coding systems.

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5. We were unable to match 96 ZIP codes to 2000 ZCTAs. These nonmatched ZIP codes are composed of single buildings and P.O. boxes, constituting at most 200 retail establishments in total (less than 1% of all retail establishments in NYC). Therefore, we do not miss a great deal of retail activity in these nonmatched ZIP codes. In addition, most of these ZIP codes are located in midtown Manhattan, and all our analyses are robust to analyses without midtown neighborhoods.

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6. The size measure is a weighted average, using the midpoint of each size category multiplied by the number of establishments in the category. For the largest category, 1,000 or more employees, we use 1,000 as the average number. Very few establishments in New York City fall into this category, so any noise introduced by this approximation is likely to be small.

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7. We average across the 10 years, because individual ZIP–industry–year statistics can be noisy and actually change slowly on a year-to-year basis.

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8. In this article, we use the term *bodegas* to refer to small stores that carry a limited selection of groceries, such as canned and frozen goods, prepackaged basic dry goods, milk, sodas and beer, and household or personal items, such as cleaning supplies and toiletries. Some also offer limited selections of fresh produce, fresh baked goods, coffee, and deli sandwiches made to order.

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9. We ran similar analyses based on educational attainment and got results substantially similar to the income analysis. Results are available on request from authors.

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10. The 2010 decennial census has not yet been released and the American Community Survey will not report income at the ZIP-code level until releasing 5-year averages in late 2010.

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11. Our related research using a longitudinal establishment database, the NETS data set, confirms that establishments do change their NAICS classification over time, although overall numbers of these changes are fairly small.

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12. All the results described in this section are robust to analyses that (a) exclude predominantly office- and retail-occupied midtown Manhattan ZIP codes, (b) exclude all Manhattan ZIP codes, and (c) exclude ZIP codes with low populations (less than 200) and low retail activity (less than 50 establishments).

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13. All the results comparing low- and moderate-/high-income neighborhoods, as defined above, are robust to analyses comparing retail metrics across neighborhoods with (a) income below and above the citywide median income and (b) share of college-educated residents above and below the share of college-educated residents for the entire city.

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14. Both variables are constructed using NYC Department of City Planning PLUTO Database for 2006. Proximity to rail and subway transit is created using GIS maps of rail and subway entrances and calculating the average distance between the entrances and retail properties for each ZIP code. The amount of retail space per building is calculated by dividing the aggregate amount of retail space for the ZIP code by the total number of commercial and mixed-use buildings in the ZIP code.

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15. All the results for the dynamic analysis are robust to analyses that (a) exclude predominantly office- and retail-occupied midtown Manhattan ZIP codes, (b) exclude ZIP codes with low populations (<200) and low retail activity (<50 establishments), and (c) other thresholds for low- and moderate-/high-valued neighborhoods (specifically 60%, 10%, and 120% of the average price per unit for New York City overall).

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16. We cannot use the chains or supermarket data in the dynamic analysis, because we have those data for only one point in time (2009).

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17. [Schuetz, Kolko, and Meltzer \(2010\)](#) conducted a large-scale analysis of retail and neighborhood income for 58 large U.S. metro areas. They found that retail density varies with income for certain retail types, such as food service and chain supermarkets and drugstores. In addition, average establishment size increases with income for all retail types. Retail density increases with population density, as expected, and decreases with distance to CBD and with share of owner-occupied housing.

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Appendix

Lots with > 200 sq ft of Retail

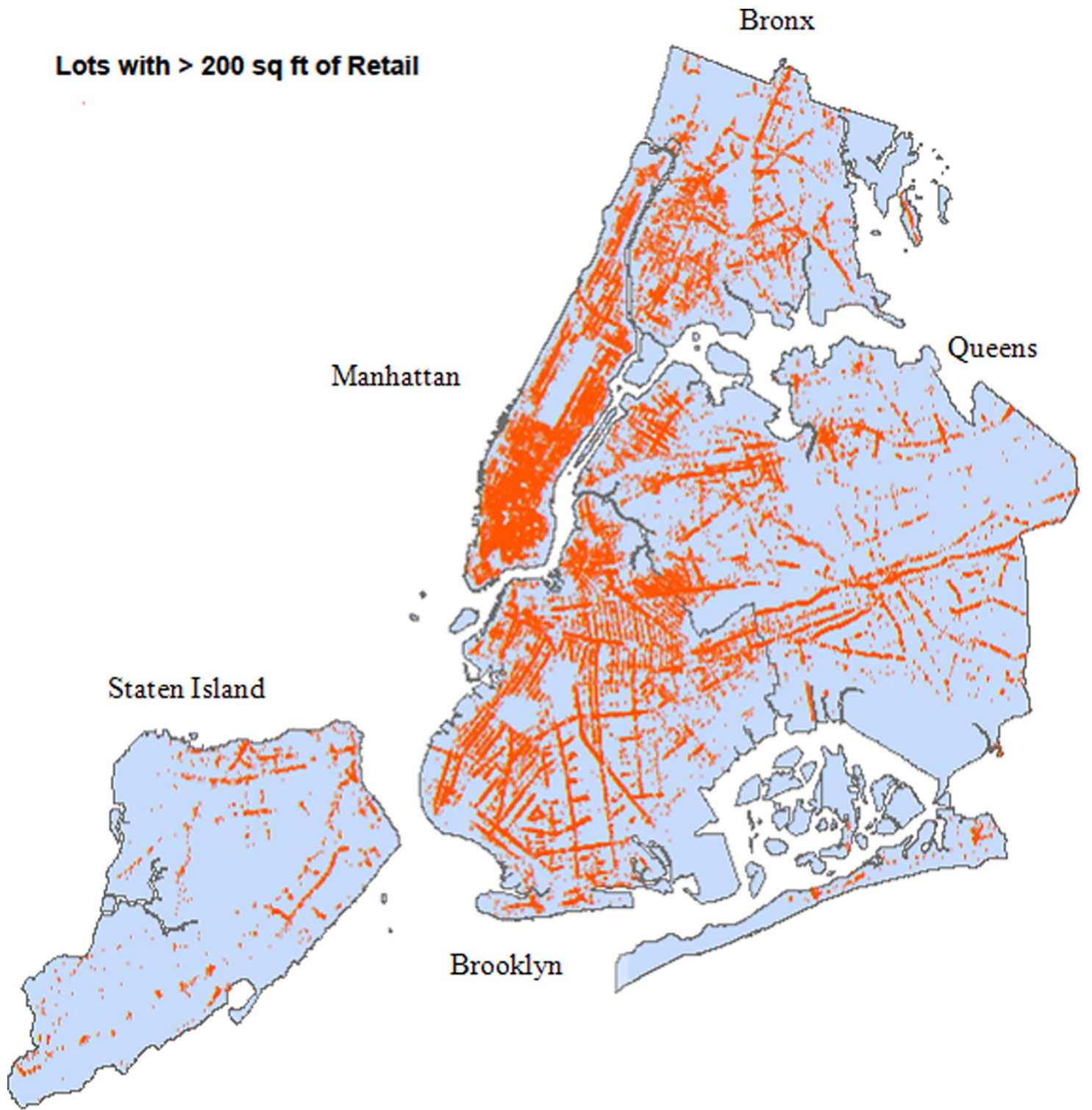


Figure A1. Retail corridors, New York City

Table A1. Chain Stores and Restaurants in Database

Category	Chain name
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Category	Chain name
Clothes and shoes	American Apparel, Ann Taylor, Banana Republic, Brooklyn Industries, Foot Locker, Gap, H&M, Marshalls, Old Navy, Payless, The Childrens Place, Urban Outfitters
Drugstore	CVS, Duane Reade, Rite Aid, Walgreens
Financial services	H&R Block, Jackson Hewitt, Liberty Tax
Food/beverage: Fast food	Auntie Annies, ^a Burger King, ^a Crown Fried Chicken, ^a Dominos, ^a Five Guys, ^a Golden Krust, ^a KFC, ^a Master Wok, ^a Mcdonalds, ^a Nathans, ^a Papa Johns, ^a Pizza Hut, ^a Popeyes, ^a Pretzel Time, ^a Ranch1, Sbarro, ^a Taco Bell, ^a Wendys, ^a White Castle ^a
Food/beverage: Other	7-Eleven, Applebee's, Arthur Treacher, Au Bon Pain, Baskin Robbins, Ben and Jerrys, Blimpie, Boston Market, Carvel, Chevy's, Chipotle, Chuck E. Cheese, Cold Stone Creamery, Cosi, Crumbs, Dallas BBQ, Dunkin Donuts, Famiglia, Fridays, Gloria Jeans Coffee, Godiva, Haagen-Dazs, Hale and Hearty, Hard Rock Café, Ihop, Jamba Juice, Johnny Rockets, Juan Valdez, Le Pain Quotidien, Mrs Fields, Olive Garden, Outback, Panera Bread, Pax Wholesome, Pinkberry, Pret a Manger, Quiznos, Red Lobster, Starbucks, Subway, Tasti D-Lite, Tim Horton, Two Boots, UNO'S
Gyms	Bally's Total Fitness, Crunch, Curves, David Barton, Equinox, Gold's Gym, Lucille Roberts, NY Sports Club
Home goods	Home Depot, Rent-a-Center

Source. Adapted from [Center for an Urban Future \(2009\)](#).

a. "Unhealthy" fast-food restaurant.

Table A2. Supermarket Chains in Database

Supermarket name
Associated Supermarkets
Bravo
Citarella ^a

Supermarket name

Costco

C-Town

D'Agostino^a

Fairway

Fine Fare

Food Emporium

Garden of Eden^a

Gourmet Garage^a

Gristedes

Key Food

Morton Williams

Pathmark

Trader Joe's

West Side Market

Western Beef

Western Beef/Junior's

Whole Foods^a

Source. Adapted from [Center for an Urban Future \(2009\)](#), additional online research by authors.

a.

"Upscale" supermarket.

Table A3. Does Retail Access Improve in High-Value Neighborhoods That Upgrade?

Percentage Change in Housing Values (1998-2007)

	Upgrading	Stable/Lagging	Difference
Percentage change est/acre			
Retail (44)	2.96	1.86	1.10
Food service (72)	29.70	17.35	12.34**
Groceries	4.57	8.11	-3.54**
Clothing	14.98	-7.38	22.36*
Percentage change emp/acre			
Retail (44)	35.07	9.15	25.92***
Food service (72)	40.06	23.86	16.20**
Groceries	41.94	10.69	31.25**
Clothing	47.75	14.95	32.80*
Percentage change emp/est			
Retail (44)	32.66	7.39	25.27***
Food service (72)	11.16	6.92	4.24
Groceries	37.84	3.11	34.73***
Clothing	35.62	22.18	13.44
Percentage change Herfindahl			
Retail (44)	71.65	78.48	-6.83
Food service (72)	-4.68	3.32	-7.99**
<i>n</i>	18	58	

Note. All ZIPs had initial housing values (1998) $\geq 80\%$ NYC average “Upgrading” defined as ZIP percentage change in average housing value $>$ NYC percentage change in average housing value (1998-2007). Stable/lagging defined as ZIP percentage change in average housing value \leq NYC percentage change in average housing value (1998-2007).

*

$p < .10$. ** $p < .05$. *** $p < .01$.

References

Alwitt L., Donley T. (1997). Retail stores in poor neighborhoods. *Journal of Consumer Affairs*, 31, 139-164.

[+ Show References](#)

[Google Scholar](#)

Barragan R. (2010, September 9). *A one stop shop for small and micro business*. Presentation at Federal Reserve Bank of Kansas City, Economic Development in Underserved Communities Conference, Kansas City.

[Go To Reference](#)

[Google Scholar](#)

Berry B. (1967). *Geography of market centers and retail distribution*. Englewood Cliffs, NJ: Prentice Hall.

[Go To Reference](#)

[Google Scholar](#)

Berry S. T., Waldfogel J. (2003, May). *Product quality and market size*. (NBER working paper No. 9675). Cambridge, MA: National Bureau of Economic Research.

[Go To Reference](#)

[Google Scholar](#)

Block J., Scribner R., DeSalvo K. (2004). Fast food, race/ethnicity, and income: A geographic analysis. *American Journal of Preventive Medicine*, 27, 211-217.

[Go To Reference](#)

[Google Scholar](#)

Bondonio D. (2003). *Do tax incentives affect local economic growth? What mean impacts miss in the analysis of Enterprise Zone policies* (WP 03-17). Washington, DC: U.S. Census Bureau, Center for

[Go To Reference](#)

[Google Scholar](#)

Busso M., Kline P. (2008, February). *Do local economic development programs work? Evidence from the Federal Empowerment Zone Program* (Cowles Foundation Discussion Paper No. 1638; Yale Economics Department Working Paper No. 36). Retrieved from http://papers.ssrn.com/sol3/papers.cfm?abstract_id=1090838

[Go To Reference](#)

[Google Scholar](#)

Caplovitz D. (1967). *The poor pay more: Consumer practices of low-income families*. New York, NY: Free Press.

[Go To Reference](#)

[Google Scholar](#)

Carr J., Schuetz J. (2001). *Financial services in distressed communities: Framing the issue, finding solutions*. Washington, DC: Fannie Mae Foundation.

[Go To Reference](#)

[Google Scholar](#)

Case A., Katz L. (1991, May). *The company you keep: The effect of family and neighborhood on disadvantaged youth* (NBER Working Paper No. 3705). Cambridge, MA: National Bureau of Economic Research.

[Go To Reference](#)

[Google Scholar](#)

Center for an Urban Future. (2009). *Return of the chains*. Retrieved from http://www.nycfuture.org/images_pdfs/pdfs/ReturnoftheChains.pdf

[+ Show References](#)

[Google Scholar](#)

Chapple K., Jacobus R. (2009). Retail trade as a route to neighborhood revitalization. In Wial H., Pindus N., Wolman H. (Eds.), *Urban and regional policy and its effects* (Vol. 2, pp. 19-68). Washington, DC: Brookings Institution/Urban Institute.

[Go To Reference](#)

[Google Scholar](#)

Coopers & Lybrand Economic Studies Group. (1982). *Impact of enterprise zone tax incentives on selected small businesses*. Washington, DC: Author.

[Go To Reference](#)

[Google Scholar](#)

Cutler D., Glaeser E., Vigdor J. (1999). The rise and decline of the American ghetto. *Journal of Political Economy*, 107, 455-506.

[Go To Reference](#)

[Google Scholar](#)

Davis S., Haltiwanger J., Schuh S. (1996). *Job creation and destruction*. Cambridge: MIT Press.

[Go To Reference](#)

[Google Scholar](#)

DiPasquale D., Wheaton W. (1996). *Urban economics and real estate markets*. Boston, MA: Prentice Hall.

[Go To Reference](#)

[Google Scholar](#)

Dowall D. (1996). An evaluation of California's enterprise zone programs. *Economic Development Quarterly*, 10, 352-368.

[Go To Reference](#)

[Google Scholar](#)

Galster G., Walker C., Hayes C., Boxall P., Johnson J. (2004). Measuring the impact of Community Development Block Grant spending on urban neighborhoods. *Housing Policy Debate*, 15, 903-934.

[Go To Reference](#)

[Google Scholar](#)

Glaeser E., Gottlieb J. (2006). Urban resurgence and the consumer city. *Urban Studies*, 43, 1275-1299.

[Go To Reference](#)

[Google Scholar](#)

Glaeser E., Kolko J., Saiz A. (2001). Consumer city. *Journal of Economic Geography*, 1, 27-50.

[Go To Reference](#)

[Google Scholar](#)

Glickman N. J. (1981). Emerging urban policies in a slow-growth economy: Conservative initiatives and progressive responses in the U.S. *International Journal of Urban and Regional Research*, 5, 492-527.

[Go To Reference](#)

[Google Scholar](#)

Haltiwanger J., Jarmin R., Krizan C. J. (2010). Mom-and-pop meet Big-Box: Complements or substitutes? *Journal of Urban Economics*, 67, 116-134.

[Go To Reference](#)

[Google Scholar](#)

Hayes L. R. (2000). *Do the poor pay more? An empirical investigation of price dispersion in food retailing* (Industrial Relations Working Paper No. 446). Princeton, NJ: Princeton Department of Economics.

[Go To Reference](#)

[Google Scholar](#)

Hebert S., Vidal A., Mills G., James F., Gruenstein D. (2001). *Interim assessment of the Empowerment Zones and Enterprise Communities (EZ/EC) Program: A progress report*. Retrieved from http://www.huduser.org/Publications/pdf/ezec_report.pdf

[Go To Reference](#)

[Google Scholar](#)

Helling A., Sawicki D. (2003). Race and residential accessibility to shopping and services. *Housing Policy Debate*, 14, 69-84.

[Go To Reference](#)

[Google Scholar](#)

Hotelling H. (1929). Stability in competition. *Economic Journal*, 39, 41-57.

[Go To Reference](#)

[Google Scholar](#)

International Council of Shopping Centers. (2004). *Developing successful retail in underserved urban markets*. New York, NY: Author.

[+ Show References](#)

[Google Scholar](#)

Jargowsky P. (2003). *Poverty and place: Ghettoes, barrios and the American City*. New York, NY: Russell Sage.

[Go To Reference](#)

[Google Scholar](#)

Kaufman P. R., MacDonald J. M., Lutz S. M., Smallwood D. M. (1997, December). *Do the poor pay more for food? Item selection and price differences affect low-income household food costs* (Agricultural Economics Report No. 759). Washington, DC: United States Department of Agriculture, Economic Research Service.

[Go To Reference](#)

[Google Scholar](#)

Lewis L. B., Sloane D. C., Nascimento L. M., Diamant A. L., Guinyard J. J., Yancey A., Flynn G. (2005). African Americans' access to healthy food options in South Los Angeles restaurants. *American Journal of Public Health*, 95, 668-673.

[+ Show References](#)

[Google Scholar](#)

Massey D., Denton N. (1993). *American apartheid*. Cambridge, MA: Harvard University Press.

[Go To Reference](#)

[Google Scholar](#)

Neal Z. (2006). Culinary deserts, gastronomic oases: A classification of US cities. *Urban Studies*, 43, 1-21.

[Go To Reference](#)

[Google Scholar](#)

Neumark D., Kolko J. (2008, December). *Do enterprise zones create jobs? Evidence from California's Enterprise Zone Program* (NBER Working Paper No. 14530). Cambridge, MA: National Bureau of Economic Research.

[Go To Reference](#)

[Google Scholar](#)

Pearce J., Blakely T., Witten K., Bartie P. (2007). Neighborhood deprivation and access to fast food retailing: A national study. *American Journal of Preventive Medicine*, 32, 375-382.

[Go To Reference](#)

[Google Scholar](#)

Pillsbury R. (2010). From Hamburger Alley to Hedgerose Heights: Toward a model of restaurant location dynamics. *The Professional Geographer*, 39, 326-344.

[Go To Reference](#)

[Google Scholar](#)

Powell L., Slater S., Mirtcheva D., Bao Y., Chaloupka F. J. (2007). Food store availability and neighborhood characteristics in the United States. *Preventive Medicine*, 44, 189-195.

[+ Show References](#)

[Google Scholar](#)

Schuetz J., Kolko J., Meltzer R. (2010, September). *Is the "shop around the corner" a luxury or a nuisance? The relationship between income and neighborhood retail patterns* (Working paper). Retrieved from http://papers.ssrn.com/sol3/papers.cfm?abstract_id=1681734

[Go To Reference](#)

[Google Scholar](#)

Waldfoegel J. (2008). The median voter and the median consumer: Local private goods and population composition. *Journal of Urban Economics*, 63, 567-582.

[Go To Reference](#)

[Google Scholar](#)

Walker C., Hayes C., Galster G., Boxall P., Johnson J. (2002). *The impact of CDBG spending on urban neighborhoods*. Washington, DC: Urban Institute. Retrieved from http://www.urban.org/uploadedpdf/410664_CDBG_Spending.pdf

[Go To Reference](#)

[Google Scholar](#)

en W. (1999). Consumption and the problem of variety: Cultural omnivorousness, social distinction and dining out. *Sociology*, 33, 105-127.

[Go To Reference](#)

[Google Scholar](#)

Wilson W. J. (1987). *The truly disadvantaged*. Chicago, IL: University of Chicago Press.

[Go To Reference](#)

[Google Scholar](#)

Zenk S., Schulz A. J., Israel B. A., James S. A., Bao S., Wilson M. (2005). Neighborhood racial composition, neighborhood poverty, and the spatial accessibility of supermarkets in Metropolitan Detroit. *American Journal of Public Health*, 95, 660-667.

[+ Show References](#)

[Google Scholar](#)

Zukin S., Trujillo V., Frase P., Jackson D., Recuber T., Walker A. (2009). New retail capital and neighborhood change: Boutiques and gentrification in New York City. *City & Community*, 8, 47-64.

[Go To Reference](#)

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