



The Effect of Community Gardens on Neighboring Property Values

Ioan Voicu* and Vicki Been**

Cities across the United States that have considerable vacant land are debating whether to foster community gardens on that land, while cities with land shortages are debating when to replace gardens with other uses. Meanwhile, many cities are looking for new ways to finance green spaces. Little empirical evidence about the neighborhood impacts of community gardens is available, however, to inform the debate or to help cities design financing schemes. This article estimates the impact of community gardens on neighborhood property values, using rich data for New York City and a difference-in-difference specification of a hedonic regression model. We find that gardens have significant positive effects, especially in the poorest neighborhoods. Higher-quality gardens have the greatest positive impact.

Introduction

In recent years, controversies have erupted in many communities, perhaps most notably New York City, about the use of vacant lots for community gardens. The gardens often are initially warmly welcomed by communities as catalysts for the improvement of troubled neighborhoods, and they frequently become treasured by neighborhood residents. When private owners or public agencies then seek to reclaim the lots for development, support for the gardens often is pitted against other social goals such as affordable housing or other public uses.

In such controversies, advocates for the gardens have advanced many arguments about the value of the community gardens, claiming that gardens stabilize and improve their host neighborhoods, provide a focal point for community organizing and social networks, bring fresh produce to neighborhood where fresh fruits and vegetables often are not available and provide recreation and exercise for neighborhood residents. Advocates have cited anecdotal evidence that gardens increased the value of neighboring properties and spurred neighborhood revitalization. As proof of the value of such spaces, advocates also have pointed

*Office of the Comptroller of the Currency, Washington, DC 20219 or Ioan.Voicu@occ.treas.gov.

**New York University School of Law, New York, NY 10012 or vicki.been@nyu.edu.

to efforts of several cities to develop small “pocket” parks through initiatives such as New York City’s “Green Streets” program.

Little empirical evidence about the effect community gardens have on surrounding properties is available, however, to inform the debates over whether to foster the development of community gardens or when to replace community gardens with other uses. That gap is surprising, given that the benefits gardens and small urban green spaces provide, relative to alternative uses of the property, have played a key role in controversies over community gardens not just in New York City, but in Boston, St. Louis and many other communities.¹ The value of gardens and small parks also is central to debates over the desirability of efforts to reduce sprawl by fostering infill development. More generally, local governments across the country face increasing controversy about how much and what types of green space and park land should be provided by the community or by developers proposing to build within a community. Further, many neighborhoods and cities are debating whether to charge developers impact fees to fund the provision of gardens and small parks within a development. In many states, the law requires that any such charges be based upon realistic assessments of the need for and value of such spaces. Similarly, local governments need reliable data about the economic impact of different kinds of parks and open space in order to decide whether to employ tax increment financing (TIF) to support the provision of parks and open space.²

Despite this critical need for information, relatively little is known about the economic value of gardens, small urban parks and other such green spaces. A number of studies have investigated the link between proximity to parks or other open space and property values, but because they focused on relatively low-density development, the results of those studies may not be transferable to the setting in which many community gardens or small urban parks are found. Further, the existing studies share methodological and data limitations that make it difficult to pinpoint the direction of the causality of any property value differences found.

¹ In addition, cities like Berkeley, Chicago, Madison, Philadelphia and Seattle have grappled with how to design programs to promote community gardens and neighborhood parks.

² Under a TIF plan, a community designates a geographic area that is likely to benefit from the provision of open space, issues bonds to finance the purchase of that space, then pays for the debt service on the bonds from the additional property tax revenues resulting from the increased value of properties within the TIF district. For a TIF to be successful, the park or other amenity financed must spur private development and raise property values sufficiently to generate the increased tax receipts necessary for the debt service. Being able to accurately predict the property value impacts that different kinds of open space will have on neighboring property values accordingly is crucial to local governments considering TIF.

In this article, we attempt to fill these gaps, using unique data from New York City and a difference-in-differences specification of a hedonic regression model to estimate the effect of community gardens on neighborhood residential property values.³ Impacts are estimated as the difference between property values in the vicinity of garden sites before and after a garden opens relative to price changes of comparable properties farther away, but still in the same neighborhood. We examine whether effects change over time or vary with neighborhood type and garden quality.

We find that the opening of a community garden has a statistically significant positive impact on the sales prices of properties within 1,000 feet of the garden and that the impact increases over time. Higher-quality gardens have the greatest positive impact. We also find that gardens have the greatest impact in the most disadvantaged neighborhoods. Finally, a simple cost–benefit analysis suggests that the gain in tax revenue generated by community gardens in the 1,000-foot ring may be substantial.

The article is organized as follows. The next section offers a review of relevant literature. The third section describes the models and empirical strategy. In the fourth section, we provide a description of the data, and in the fifth section we present results. The article ends with a summary of the key findings.

Existing Literature

Although we know of only one other study of the impact that community gardens have on surrounding properties, a number of studies have investigated the link between proximity to parks and other open spaces and property values. These studies vary widely in their methodology, ranging from simple surveys to hedonic modeling or matched-pair analysis. They have examined the influence of a wide range of open space types, such as golf courses, greenbelts, wetlands, agricultural spaces, urban parks and playgrounds. The studies that focus directly on parks of various kinds tend to show positive property value impacts. Crompton (2001) surveys the older literature, while Hobden, Laughton and Morgan (2004) summarize more recent literature. We review here the studies that focused on the small urban parks or other open spaces most analogous to community gardens.

Bolitzer and Netusil (2000) studied the effect of proximity to an open space (public and private parks, cemeteries and golf courses) in Portland, Oregon,

³ In preliminary research, we also investigated the effect of community gardens on neighboring commercial property values. Findings suggest that gardens do not have any significant immediate impacts on neighboring commercial property values, although some benefits may occur over time.

using data on 16,402 sales of single-family homes between 1990 and 1992. Using a linear hedonic model, they found that a home located within 1,500 feet ($7\frac{1}{2}$ blocks) of a 20-acre open space (the mean for the public parks in the area) sold for approximately \$2,670 (in 1990 dollars) more than homes that were farther from a park. A semi-log model showed that proximity to any open space increased a home's sales price by 1.43%. When the authors estimated the effect of the different types of open space in the study area, however, the increase in value was limited to public parks and cemeteries; private parks had no statistically significant effect on home prices. Because the mean size of the public parks studied was 20 acres, while the mean size of the private parks was almost four acres, the finding that private parks did not have a statistically significant effect on neighboring properties is most relevant for our study of community gardens, which have a median size of 6,000 square feet (less than $\frac{1}{4}$ acre) in New York City.

Lutzenhiser and Netusil (2001) refined the approach taken in Bolitzer and Netusil (2000), using the same data from Portland, Oregon. Breaking the types of open space into five categories, they found that "urban parks" (those in which more than 50% of the area was landscaped or developed for uses such as swimming pools or ball fields) had a statistically significant effect of \$1,214 (in 1990 dollars), or 1.8% of the mean house value, on the value of single-family residences within 1,500 feet of the park. That effect was far lower than the effect of both "natural area parks" in which a majority of the land was preserved in natural vegetation, and golf courses (\$10,648 and \$8,849, respectively). The authors estimated that the size of an urban park that would maximize its positive effect on a home's sales price would be 148 acres, almost 600 times larger than the median size of the community gardens in our study.

Espey and Owasu-Edusei (2001) used hedonic methods to investigate the effect proximity to 24 neighborhood parks had on the sales prices of single-family homes in Greenville, South Carolina, between 1990 and 1999. The authors broke the parks down into four categories. Type 1 parks ranged from 15,620 to 87,687 square feet and were essentially playgrounds with some grassy areas, but were not "particularly attractive." Type 2 parks were small, attractive parks with playgrounds. Type 3 consisted of attractive medium-sized parks with both sports fields or courts and playgrounds but also some natural areas; and Type 4 parks were unattractive medium-sized parks with few amenities and no natural area. Using a semi-log model, the authors found that the Type 1 parks had a statistically significant negative effect on the sales prices of homes within 300 feet of the park, a significant positive effect of about 15% on the sales of houses between 300 and 500 feet and a significant positive effect of about 6.5% on the sales prices of homes located between 500 and 1,500 feet of the park. Small

attractive parks (Type 2) had a statistically significant positive effect of 11% on the sales prices of houses within 600 feet of the park, but no statistically significant effect beyond that.

Pincetl *et al.* (2003) studied a neighborhood five miles from downtown Los Angeles in which most of the housing was multifamily, rental buildings. The neighborhood had no parks, but the authors examined the effect of greenery (as measured using aerial photographs) on 260 sales of single-family homes over an 18-month period. The authors found that an 11% increase in the amount of greenery (equivalent to a one-third acre garden or park) within a radius of 200 to 500 feet from the house increased the sales price of the house by approximately 1.5%.⁴

Hobden, Laughton and Morgan (2004) used a matched pairs methodology to measure the effect greenways had on adjacent properties. The authors grouped greenways into eight categories, based upon improvements to the greenways, and also distinguished between “small parks” in which the greenway was a narrow strip (often a pathway) of less than 50% of the area of the adjacent residential property. Using data from the years between 1980 and 2001, the authors identified 755 matched pairs. They found that adding a greenway increased the sales price of adjacent properties by 2.8%. Where the greenway was defined as a “small park,” the greenway increased the values of adjacent properties by 6.9%.

New Yorkers for Parks and Ernst & Young (2003) found, through case study methodology,⁵ that the values of single-family homes located near three well-improved parks in Brooklyn, Queens and Staten Island were 8 to 30% higher than values of homes farther from the parks.

Most directly relevant to our research, Tranel and Handlin (2006) used census data for 1990 and 2000 and a difference-in-difference methodology to assess the neighborhood effects of 54 community gardens in St. Louis, Missouri. They found that median rent, median housing costs (mortgage payments, maintenance costs and taxes) for owner-occupied housing as well as the homeownership rate increased in the immediate vicinity of gardens relative to the surrounding census tracts, following garden opening.

⁴ A recent working paper by Wachter and Wong (2006) focuses on tree plantings in Philadelphia and finds that they have positive and significant impacts on surrounding property values.

⁵ Specifically, for each case study park, the authors compare the average house price and price appreciation in the immediate neighborhood of the park with those farther away, in designated control areas, during the period of capital improvements.

The usefulness of the existing hedonic studies to many local governments is limited in several ways.⁶ First, most of the studies focus on the impacts of open spaces on single-family house values and other neighborhood indicators in relatively low-density settings that do not characterize the denser urban neighborhoods in which many controversies about establishing or preserving parks arise. See, for example, Bolitzer and Netusil (2000), Espey and Owasu-Edusei (2001), Lutzenhiser and Netusil (2001) and Tranel and Handlin (2006).

Second, the existing literature does not attempt, except in the most general way, to account for differences in the quality of the parks or gardens. See, for example, Espey and Owasu-Edusei (2001), which distinguishes between unattractive and attractive parks.

Moreover, in most previous studies, data limitations make it difficult to pinpoint the direction of causality. The existing studies usually employ cross-sectional techniques that compare sales prices and other neighborhood indicators in neighborhoods with open spaces to those in neighborhoods without, but it is difficult to know whether the two groups of neighborhoods truly are comparable. Therefore, their results could be interpreted to suggest that open spaces lead to improvements in the surrounding neighborhood, but could instead mean that such spaces are systematically located in strong neighborhoods. One exception is Tranel and Handlin (2006) who employ a difference-in-difference methodology to study the neighborhood impacts of community gardens. However, when estimating the impact the gardens have on rents and owner-occupied housing costs, they do not control for housing characteristics, thus raising questions about the comparability of the treatment and control areas. Moreover, their analysis is based on only two data points in time, one before and one after garden formation, and thus, it does not control for baseline (preexisting) trends, nor does it estimate the variation of garden impacts over time.

Finally, the existing literature does not address whether differences in the characteristics of the surrounding neighborhood affect the property value impacts of the parks or gardens.

Methodology

Our main empirical goal is to estimate the impacts of community gardens on neighboring residential property values. In this section, we outline various

⁶ Some studies attempt to address these limitations using case studies or other methodological approaches. See, for example, New Yorkers for Parks and Ernst & Young (2003). We discuss here only those studies using hedonic methods.

hypotheses about how gardens might affect neighborhood property values and describe our empirical strategy to identify the property value impacts of gardens.

Hypotheses About How Gardens Might Affect Neighborhood Property Values

In principle, gardens, parks and other open spaces could have both positive and negative impacts on surrounding communities. They may benefit neighbors by providing access to recreational facilities, fostering a sense of community, providing attractive scenery and views for nearby residents and passersby, replacing an eyesore or a location for undesirable behaviors and improving both the environmental quality and the general quality of life in the neighborhood. On the other hand, they may provide a haven for loitering or other undesirable social behavior or cause noise and congestion.⁷ If the garden is not well maintained, it may be an eyesore. As housing is fixed in space, and the value of a home is therefore influenced not only by its structural features and quality but also by its surroundings, these amenities and disamenities associated with gardens could be capitalized into the value of surrounding properties.

While garden impacts—both positive and negative—may be largest in adjacent properties, they may well extend some distance, with diminishing magnitude. However, it is not clear, a priori, how the net effect will vary with distance.

Garden effects also may vary with the garden area. Specifically, both the positive and negative externalities are likely to increase with garden area (*e.g.*, larger gardens may provide more recreational facilities but also may be noisier). Thus, it is uncertain, a priori, whether size will have a positive or negative influence on the net effect of the garden.

Additionally, garden impacts may vary over time. For example, benefits may grow over time as more neighborhood residents become familiar with the garden and get involved in various activities taking place at the garden.

The impact of a garden may vary across neighborhoods with different characteristics. As pointed out in several prior studies (Francis, Cashdan and Paxson 1984, Harnick 2000, Saldivar-Tanaka and Krasny 2004), gardens in poor neighborhoods provide an affordable alternative to city parks, which often are located in more affluent neighborhoods and are not easily accessible to poor residents. Therefore, gardens may be more valuable to these residents, leading one to expect larger positive impacts (or smaller negative impacts) on residential property values in lower-income areas.

⁷ Note, however, that some studies claim that urban gardening may, in fact, reduce local crime (Warner and Hansi 1987, Hynes 1996, Murphy 1999).

Garden effects also may vary with the housing tenure of the neighborhood residents. Homeowners may have stronger economic incentives than renters to organize and garner financial support for the initial development of the garden and its subsequent maintenance and, in general, to maintain the garden well. This would likely result in higher-quality gardens and, assuming that property value impacts would be positively related to the quality of gardens, thus lead to larger positive impacts on residential property values in neighborhoods dominated by homeowners. Additionally, homeowners may tend to be more involved in community-oriented activities—like those hosted by community gardens—than renters, either because of the homeowners' financial stake in neighborhood quality or because homeowners tend to stay in their homes for a longer period of time than do renters. On the other hand, homeowners actually may be less interested in community gardens because it is more likely that their properties already include some kind of private yard that fulfills at least some of the functions gardens serve. Because homeowners typically earn higher incomes than renters, they also may have greater access than renters to alternative open spaces and recreational opportunities. Thus, it is not clear a priori what to expect about how garden effects will vary across neighborhoods with different homeownership rates.

Because renters typically earn lower incomes than owners and live in poorer neighborhoods, it is difficult to empirically disentangle the variation in garden impacts along each of the specific neighborhood characteristics outlined above. Thus, as described in more detail below, we only differentiate impacts according to the neighborhood income level, but interpret any differences we may find as the net result of the interplay between income and tenure.

Baseline Model

Identifying the neighborhood impacts of community gardens is challenging, primarily because garden sites might not be randomly chosen (gardens might, for example, be sited on property with the lowest value or with the least development potential or might be sited in neighborhoods with activists or entrepreneurs who are making other improvements as well). Our basic strategy to address those challenges relies on a hedonic regression model with a difference-in-difference specification. Hedonic regression models explain the sales price of a property as a function of its structural characteristics (such as lot size and building age) and its neighborhood surroundings. In brief, we compare the sales prices of properties that are within designated distances—such as 1,000 feet—of community gardens to prices of comparable properties that are outside the designated ring, but still located in the same neighborhood (defined here as a

census tract).⁸ Then we compare the magnitude of this difference before and after the garden is opened.⁹ This difference-in-difference in property values is our measure of the impact gardens have on neighborhoods. The difference-in-difference measure avoids having to compare properties near gardens to other properties in different neighborhoods and accordingly avoids bias that might be introduced by any systematic differences between neighborhoods that host community gardens and other locations around the city.

To be concrete, our basic model takes the following form:

$$\begin{aligned} \ln P_{icdt} = & \alpha + \beta X_{it} + \delta_c W_c + \gamma_{dt} I_{dt} + \rho_1 \text{InRing}_i \\ & + \rho_2 \text{InRing}_i * D_i + \rho_3 \text{PostRing}_{it} + \rho_4 \text{PostRing}_{it} * D_{it} \\ & + \rho_5 \text{PostRing_GArea}_{it} + \rho_6 \text{PostRing_GArea}_{it}^2 \\ & + \rho_7 \text{PostRing_GArea}_{it} * D_{it} + \rho_8 \text{PostRing_SharePvtOwn}_{it} \\ & + \rho_9 \text{TPost}_{it} + \rho_{10} \text{TPost}_{it}^2 + \rho_{11} \text{TPost}_{it}^3 + \rho_{12} \text{TPost}_{it} * D_{it} + \varepsilon_{it}, \quad (1) \end{aligned}$$

where $\ln P_{icdt}$ is the log of the per-unit sales price of property i in census tract c , in community district d and in quarter t ; X_{it} is a vector of property-related characteristics, including age, size and structural characteristics; W_c are a series of census tract fixed effects, which help control for unobserved, time-invariant features of different neighborhoods; I_{dt} are a series of dummy variables indicating the quarter and community district¹⁰ of the sale, which

⁸ A typical city block (North/South blocks between streets) in Midtown Manhattan is about 260 feet long. Thus, the 1,000-foot ring allows for impacts extending up to roughly four blocks away from the garden. The 1,000-foot radius was selected following Bolitzer and Netusil (2001) and Netusil (2005). The former assumed, based on discussions with park officials, that the impacts of urban parks may extend to up to 1,500 feet; the latter used a 1,300-foot (1/4 mile) radius to capture impacts of parks and other relatively large open areas (golf courses, wetlands, etc.). Because New York City community gardens are relatively small, we chose a somewhat smaller ring size. This choice of ring size is also supported by discussions with members of the garden community in New York. One gardener in the Bronx mentioned, for example, that the children from a school located about four blocks away from the garden did not participate in educational activities organized at the garden because they thought it was too far away.

⁹ Ellen *et al.* (2002) employ a similar methodology to evaluate the impact of investments in selected homeownership developments.

¹⁰ Community districts are political boundaries unique to New York City. The City is divided into a total of 59 community districts, each of which has a Community Board whose members are appointed by the Borough President, with half nominated by the City Council members who represent the district. The Community Boards review applications for zoning changes and make recommendations for budget priorities.

allow for distinct time trends for each of the 51 community districts used in the analysis;^{11,12} and the remaining covariates are ring variables (described more fully below) that indicate proximity to garden sites. The coefficients to be estimated are α , β , δ , γ and ρ , and ε is an error term. Because we measure sales prices as logarithms, the coefficients can be interpreted as the percentage change in price resulting from an additional unit of the independent variable. For the dummy variables, the coefficients can be interpreted approximately as the percentage difference in price between properties with the attribute—say a garage or a corner location—and those without.¹³

Our key variables of interest are the ring variables, which capture the proximity to community gardens. We include three vectors of ring variables: *InRing*, *PostRing* and *TPost*. Specifically, the *InRing* variable is a dummy variable that takes on a value of one if the property is located within 1,000 feet of the site of an existing or future garden.¹⁴ Intuitively, the coefficient on the *InRing* variable captures the baseline difference in sales prices between properties located within a 1,000-foot ring of a garden and those farther away, but still in the same neighborhood.¹⁵ We interact *InRing* with D_i , the Euclidean distance between property i and the nearest garden. This interaction term allows us to

¹¹ Most previous research has assumed that trends in housing prices are constant across a city or metropolitan area, but this seems particularly inappropriate in a city as large and diverse as New York. Schwartz, Susin and Voicu (2003), for instance, find considerable variation in price trends across community districts in New York City.

¹² While specifying the time dummies using an even smaller geographic area—say a census tract—may seem preferable to the community districts, doing so comes at a considerable cost and adds little explanatory power. Put simply, census-tract specific time dummies would add approximately 150,000 more dummy variables to the specification, significantly increasing the number of parameters to be estimated, and greatly reducing degrees of freedom. Moreover, there is little variation in the time dummies within the community districts—an F test could not reject the hypothesis that census tract-quarter dummy variables were the same within a community district.

¹³ More precisely, the coefficient on a dummy variable should be interpreted as the difference in log price between properties that have the attribute and those that do not. Because the difference in log price closely approximates the percentage difference in price when the difference is small enough and because differences discussed in this article are generally smaller than 10%, we use this more intuitive interpretation throughout the article. The percentage effect of a difference in logs, b , is given by $100(e^b - 1)$, although this formula is itself an approximation when b is a regression coefficient; see Halvorsen and Palmquist (1980) and Kennedy (1981).

¹⁴ Some sold properties are within 1,000 feet of two or more gardens, however, a majority (almost two-thirds of all sales within 1,000 feet of any garden) are within 1,000 feet of only one garden.

¹⁵ In preliminary research, we allowed baseline property values to vary also with both the size and the land ownership of the site; however, we did not find any statistically significant variations along these dimensions and settled on the more parsimonious specification shown here.

estimate how the effects of proximity to a garden site vary with distance to the site, within the 1,000-foot ring.

The *PostRing* dummy variable indicates whether the property is within 1,000 feet of an existing garden. Its coefficient provides the simplest impact estimate. The existing garden area within the ring of the property, *PostRing_GArea*, (and its square) offers a measure of the marginal effects of additional square feet of garden area. The share of existing garden area in the ring that is privately owned, *PostRing_PvtOwn*, captures differences in property prices due to differences in land ownership.¹⁶ Finally, to allow the impact to vary over time, we include a postcompletion trend variable, *Tpost*, its square and its cube.¹⁷ Specifically, *Tpost* equals the number of years between the date of sale and the garden opening date for properties in the 1,000-foot ring.¹⁸ Note that because we control for the existing garden area within the ring of a sale and the share of this garden area that is privately owned, as well as for the time elapsed since the garden opened, the *PostRing* coefficient should be viewed as the fixed impact of a garden sited on publicly owned land (independent of its size), immediately after its opening.

We also interact the *PostRing* variable with distance to allow impacts to vary with distance. In addition, by interacting distance with *PostRing_GArea* and *Tpost*, we explore how the distance gradient of impacts changes with garden area and over time.^{19,20}

¹⁶ Of the gardens we studied, 95% are sited on publicly owned parcels that are leased to local communities for gardening; the rest are located on privately owned lots. Throughout the article we use the terms “gardens sited on publicly owned (privately owned) land” and “public (private) gardens” interchangeably.

¹⁷ Preliminary investigation of the data suggested that a third degree polynomial would be desirable to capture the general postopening time trend in house prices in rings. Nonetheless, results are robust to the choice of the power terms.

¹⁸ To be clear, *Tpost* equals 1/365 if a sale is located within the ring of a garden and occurs the day after its opening, it equals one if the sale occurs one year after the garden opening and so on. The environmental disamenities literature has explored alternative ways to specify the decay or acceleration of impacts over time. See Kiel and Zabel (2001) for a useful discussion.

¹⁹ In preliminary work, we also allowed for nonlinear distance gradients by including distance squared terms; because their coefficients were not statistically significant and results are robust to their exclusion we opted to exclude them from the models shown here.

²⁰ Although not shown in Equation (1), our specification also includes a set of control variables that capture proximity to other gardens that could not be used for impact estimation either due to missing opening date or because the opening date was outside of the period covered by our sales data. If we did not include these controls and the location of these other gardens were correlated with that of our sample of gardens with valid data, our impact estimates would be biased.

Heterogeneity of Impacts Across Different Neighborhood Types

We next explore the extent to which the impacts of community gardens vary with income levels in a neighborhood. While Schwartz *et al.* (2006) as well as Santiago, Galster and Tatian (2001) find significant differences in the impact of publicly subsidized housing investments across submarkets, nobody has explored such variation in the context of green spaces, despite its relevance to such policy issues as environmental justice and the equity of municipal service provision. We test for heterogeneity in impacts between low- and high-income areas by interacting all of our ring variables, ring-distance interaction variables and hedonic variables with a dummy variable indicating neighborhood income level.²¹ Following Schwartz *et al.* (2006), we identify two submarkets—defined by community districts—based on household income information from the 1990 Decennial Census: the low-income submarket consists of community districts with an average household income less than 80% of the MSA mean household income and the higher-income submarket includes all the remaining districts.²²

Garden Quality

To better understand the effects different types of gardens may have on neighboring property values, we conducted a survey to gather qualitative information on various garden features and then extended our basic model to incorporate this information. In the survey, we focused on a subset of our sample of community gardens—those located in the Bronx. Following onsite visits, we ranked the Bronx gardens on the following criteria: accessibility to the general public, fencing attractiveness and permanence, cleanliness, landscaping quality, presence of decorations, existence of social spaces and overall condition of the garden.

For example, the accessibility rating is derived from questions about whether the garden advertises the hours during which it is open to public, whether the garden provides information about how to get involved with community gardening, and whether there are signs announcing the organization sponsoring the garden (a detailed description of survey questions is available upon request from the authors). If more than half of the answers under a given criterion reflected a favorable opinion, the garden was rated “Acceptable” on that criterion. The rating of the overall condition of the garden was derived from the whole survey questionnaire. Because no garden had less than six or more than 23 favorable answers (out of a total of 24 questions), we choose the midpoint of the (6, 23)

²¹ In earlier models, an *F*-test rejected the hypothesis that the coefficients on property characteristics are similar across neighborhoods.

²² To create submarkets, we matched census tract-level data to community districts.

interval (14) as the threshold for “Acceptable” overall condition. To test the sensitivity of our results to alternative measures of overall condition, we also asked the surveyors to provide an overall assessment of the garden, on a scale of 1 to 5. A rating of 3 or more points was considered “Acceptable.” In addition, we constructed continuous measures of overall condition, equal to the number of favorable answers in the survey questionnaire or to the number of points in surveyors’ overall assessment. Finally, we considered a “Low/Good/Excellent” quality rating instead of just “Acceptable/Unacceptable.”

Using these garden ratings, we then extend our basic model to include two additional variables—share of existing or future garden area with acceptable overall quality within the ring of the property, *InRing_ShareAQ*, and share of existing garden area with acceptable overall quality within the ring of the property, *PostRing_ShareAQ*²³—to control for the overall condition of the garden.^{24,25}

Alternative Specifications

The above model controls for preexisting differences in price levels between properties within 1,000 feet of garden sites and those further away. However, it is possible that the location of gardens might be related in some systematic way to preexisting trends in house prices. Community gardens might tend to be established, for example, in microneighborhoods that were starting to appreciate, relative to the surrounding area, even before the advent of the gardens. Many gardens are located in some of the most distressed sites in the city, and it is possible that the value of the most distressed areas was rising as prices bottomed out and private developers began to invest. Or it may be that what determines whether a garden is founded is not just the initial position of a neighborhood (*e.g.*, property value levels), but also the levels of social capital

²³ To be clear, if the property is within 1,000 feet of only one garden, the share variable takes on either value 0 (if the garden quality is unacceptable) or value 1 (if the garden quality is acceptable); if the property is within 1,000 feet of two or more gardens, the share variable takes on values between 0 and 1 (*e.g.*, if there are two equally large gardens within 1,000 feet of a given property, one of which is of acceptable quality and the other is of unacceptable quality, the share variable equals 0.5).

²⁴ In sensitivity analysis using the alternative measures of overall condition, we replace these variables either with the shares of garden area of low and excellent quality or with the number of favorable answers (or points in surveyors’ overall assessment).

²⁵ In an additional extension, we allow impacts to vary with the garden’s rating on each of the six criteria described above. For this, we replace share of existing garden area with acceptable overall quality with six variables representing the shares of existing garden area with acceptable rating on each of the criteria of community access (*PostRing_ShareACA*), fencing attractiveness/permanence (*PostRing_ShareAF*), cleanliness (*PostRing_ShareAC*), landscaping (*PostRing_ShareAL*), decorations (*PostRing_SharePD*) and social spaces (*PostRing_SharePS*).

the neighborhood enjoys. If so, a garden neighborhood might have enjoyed property value increases, even before the garden was opened, because of the fact that local community activists were pushing for beneficial changes (including but not limited to a garden). Yet a third possibility is that there might have been a change in the population preceding the garden opening (like the arrival of foreign-born residents). All these scenarios suggest that what we interpret as garden impacts might simply be a continuation of these prior trends.

To help mitigate these concerns, in an extension of our main model, we follow Santiago, Galster and Tatian (2001) and estimate a specification that includes controls for trends in the relative price of housing in the vicinity of garden sites prior to the garden opening. In this specification, the counterfactual is that the price gap between properties in the vicinity of garden sites and properties in the larger neighborhood would have continued to shrink (or grow) at the preopening rate, had no garden been built. This specification differs from Equation (1) mainly in that here we add a ring-specific time-trend that measures the overall price trend in the ring (not simply the trend after completion). This variable is defined in much the same way as T_{post} is, except that it also applies to properties sold prior to project completion. For example, if a property is sold exactly two years before garden opening, the trend takes the value of -2 . An increase in prices in the ring right before opening may have been related to the gardens themselves, however, because site cleanup and garden development typically precedes the opening of a garden by about one year.²⁶ Accordingly, to avoid capturing this garden-related price increase in the prior trend, we additionally include a dummy variable, which is equal to one if the property is sold during the year immediately preceding the garden opening.

In an additional, more flexible, specification, we divide the ring-specific time-trend into three linear segments (splines), with knot-points at 10 and six years prior to garden opening (or five years prior to the start of the site clean-up). This specification allows us to more accurately capture trends, which are contemporaneous with the garden.²⁷ In this specification, the counterfactual that the price difference between properties near gardens and those in the broader

²⁶ Experts on community gardens in New York City advised us that it takes between several months and two years to set up a community garden. Thus, one year seems a reasonable assumption for the average length of time needed to create a garden.

²⁷ Note, however, that this specification may be overly conservative if the planning of the garden started well in advance of the actual work. In this case, the anticipation of the effect the garden would have on the surrounding community may have resulted in price changes before the start of the garden development and thus, the price appreciation in the years immediately preceding this start may have been caused, at least in part, by the garden itself.

neighborhood would have continued to expand or contract regardless of the garden is based on the preopening rate of price change during the five-year period preceding the start of the site clean-up.

Another concern is that other activities, unrelated to gardens, may have been taking place in the microneighborhoods surrounding the garden sites, around the time of garden openings. Then, the change in property values that we interpret as the garden impact may be due, at least in part, to these other activities. The difference-in-difference methodology we use helps to minimize the possibility that other forces are behind any property value impacts we see because that methodology requires that such other forces would have to occur in the same microneighborhoods and be triggered at the same time as the opening of the garden to show up in our results. In addition, in other preliminary research, we found that a garden opening has little effect on neighboring commercial property values (see footnote 3), so such other forces also would have to have affected residential but not commercial properties. But to further minimize the possibility that other changes in the neighborhoods in which gardens are opened may be causing any property value impacts we attribute to the gardens, we have extended our analysis to address one of the changes most likely to have occurred in the neighborhoods in which gardens were opened.²⁸ During our study period, New York City's own capital program, the Ten-Year Plan for Housing, funded the construction or rehabilitation of over 200,000 units of affordable housing (see Schill *et al.* 2002 for a detailed description of the Ten-Year Plan programs). Additionally, about 75,000 housing units were created or rehabilitated with federal subsidies (see Ellen *et al.* 2007 for a detailed description of the federal programs). Like community gardens, this assisted housing was most often sited on blighted, city-owned plots of land (Schill *et al.* 2002, Ellen *et al.* 2007), so it is possible that assisted housing was impacting some of the same neighborhoods in which gardens were being established.

To account for these housing investments, we enhance our baseline specification with sets of ring variables (*InRingSHsg*, *PostRingSHsg*, *PostRingSHsg_Units* and *TPostSHsg*) that control for proximity to various types of subsidized

²⁸ One could also imagine that a community opening a garden might at the same time undertake a broader range of neighborhood improvement initiatives that would confound any estimates of the property value impact of the garden. However, experts on community gardens in New York City informed us that while garden communities may embark on other neighborhood improvement activities, such activities usually lag behind the formation of the gardens. Gardens become catalysts of community development, as the networks and other social capital formed over gardens are deployed to start fixing schools and housing, organizing neighborhood watches and serving other community needs.

housing.²⁹ These include city-sponsored projects and housing units sponsored through federal programs such as Public Housing, Section 8, Section 202, Low-Income Housing Tax Credit (LIHTC), Section 236 and Below Market Interest Rate (BMIR).

Finally, one may be concerned that our hedonic analysis may yield biased impact estimates if there are relevant property characteristics that are unmeasured or omitted from the regression equation. In this case in particular, it seems possible that the mix of properties that sells in the vicinity of garden sites changes after the garden opening in ways that are not captured by the hedonic variables. Perhaps the “better” properties sell once the garden is set up. If so, then hedonic analysis may overstate the spillover benefits.

To address this concern, we reestimate our baseline model using a repeat-sales methodology, which relies solely on properties that sold multiple times over the study period. The repeat-sales methodology is equivalent to including fixed effects at the level of the individual property, and thus effectively controls for any time-invariant characteristics of properties (both in terms of neighborhood features and qualities of the housing itself) that may be unobserved (or omitted) in the hedonic equations. The repeat-sales methodology may, however, be less desirable than the hedonic approach due to its inherent selection bias problems (only properties that sell multiple times are included). An additional drawback of the repeat sales approach is that the model yields no estimates of the coefficients on variables that are time-invariant or that change only rarely or slowly, such as the *InRing* variables or the fixed structural characteristics of housing.

Summary of Data

To undertake the analysis outlined above, we obtained data from the Council on the Environment of New York City (CENYC) describing all the community gardens in the city.³⁰ For each garden, this data set indicates the exact

²⁹ *InRingSHsg* indicates whether the property sold is within 1,000 feet of any existing or future subsidized housing project; *PostRingSHsg* indicates whether the property sold is within 1,000 feet of any completed subsidized units; *PostRingSHsg Units* represents the number of completed subsidized units within 1,000 feet of the sale; *TPostSHsg* equals the number of years between the date of sale and the project completion date for properties in the 1,000-foot ring.

³⁰ We obtained these data in 2004. CENYC updates the data periodically, however, different gardens may have different dates of last update depending on when CENYC last received information from the gardeners or some other source. For example, the year of last update for a majority (55%) of the gardens in our main estimation sample is 2002 and the information on the remaining gardens was updated in 2003 and 2004.

location (tax lot(s)), area, opening date and land ownership.³¹ Our main estimation sample of gardens includes 636 gardens, established between 1977 and 2000.³²

As the CENYC database does not contain any qualitative data about the gardens, we inspected a subset of the city's gardens to obtain such information, as noted above. We chose to survey the gardens in the Bronx because Brooklyn had so many gardens that it was impossible to inspect them all, and we feared that gardens in relatively affluent and high-density Manhattan might not be representative of gardens across the City. The Bronx contains a mix of housing and building types, so is more representative of the City's average neighborhood (and the average neighborhood in most large cities across the nation) than is Manhattan. As shown in Table 3A, gardens in the Bronx are smaller, on average, than all New York City gardens, and were somewhat older than all New York City gardens (44.3% of the Bronx gardens were established before 1991, compared with 38.1% of all gardens). We have no reason to believe that the relationship between the quality of the garden and its impacts on surrounding property would be different in the Bronx than in the rest of the city.

We deployed teams of students from NYU School of Law and NYU's Robert F. Wagner Graduate School of Public Service to visit these gardens and rank them on such criteria as the garden's accessibility to the general public, how well maintained the garden appears, whether there are social spaces in the garden and whether the garden appears to contain trash or other disamenities.

³¹ For a majority of the gardens, the opening date recorded in the CENYC data actually represents the date when the first lease for the site was issued by Green Thumb, the city agency in charge of leasing city-owned land at no charge to neighborhood groups for gardening. According to CENYC staff, the actual opening may have occurred either before, or, more often, after the first lease was secured. To test the sensitivity of our results to changes in the opening date, we estimated our baseline models using alternative opening dates: the date recorded in the data and the date recorded in the data \pm one year. Results differ little in response to different dates. We present in the article the results based on the date recorded in the data (the positive impact estimates are actually slightly smaller when using this date).

³² The CENYC raw data included 783 gardens. Out of these, we focused on the 636 gardens established between 1977 and 2000. The reason for this selection criterion is that our sales data only covers the period 1974–2003, and it is generally desirable to match a minimum of two to three years of property sales data both before the earliest garden opening date and after the latest opening date. This approach ensures that the estimates of both pre- and post-opening levels and trends in prices in the micro-neighborhoods around the garden sites will be representative of all gardens included in the sample. Nonetheless, we also included in the analysis the rest of 147 gardens which were established outside of the 1977–2000 interval or had missing foundation year—but only to control for proximity to them and thus to obtain accurate impact estimates for the gardens on which we focus.

We were able to obtain valid information for 86 gardens (out of the 147 Bronx gardens).³³

We supplemented our data on community gardens with geocoded data from several other sources. First, through an arrangement with the New York City Department of Finance, we obtained a confidential database that contains sales transaction prices for all apartment buildings, condominium apartments and single-family homes over the period 1974–2003.³⁴ Our sales sample includes 517,791 property sales, spread across 1,799 census tracts.³⁵ Both because of the long time span of the data and New York City's size, this is a very large sample size compared with much of the literature.

Second, data on building characteristics were obtained from an administrative data set gathered for the purpose of assessing property taxes (the RPAD file). Unfortunately, the RPAD data contain little information about the characteristics of individual units in apartment buildings (except in the case of condominiums).³⁶ Nonetheless, these building characteristics explain variations in prices surprisingly well, suggesting the data are rich enough for estimating hedonic price equations.³⁷

³³ We eliminated from the analysis 55 garden sites (37% of the Bronx total) which, upon inspection, turned out to not actually host an active community garden (*e.g.*, the site was vacant or abandoned, or hosted a building, a school playground or a school garden), as well as six active community gardens (4% of the total) for which the surveyor could not obtain reliable information on one or more items of interest. One possible reason for the high percentage of sites that turned out not to be active gardens is that the uses of some of the sites included in the CENYC database may have changed between the time of the latest data update and the date of our survey (for a large majority of these sites, the latest data update was beginning of 2002, whereas our survey took place towards the end of 2004). Because we do not know when the use of the lot changed, we did not exclude these gardens from our baseline model.

³⁴ Note that sales of cooperative apartments are not considered to be sales of real property and are not included in the DOF data set. Most of the apartment buildings in our sample are rent stabilized. Given that legally allowable rents were typically above market rents outside of affluent neighborhoods in Manhattan and Brooklyn during the period of our study, we do not believe that their inclusion biases our results (see Pollakowski 1997).

³⁵ We limited the analysis to properties that are located within the 51 community districts (of the total 59) with community gardens.

³⁶ Most of the RPAD data we use were collected in 1999, and it is conceivable that some building characteristics may have changed between the time of sale and 1999. However, most of the characteristics that we use in the regressions are fairly immutable (*e.g.*, corner location, square feet, presence of garage), and when we merged RPAD data from 1990 to 1999, we found that characteristics changed very rarely. Even among these apparent changes, we suspect that a majority are corrections, rather than true changes.

³⁷ See Ellen *et al.* (2002) for more detail on the data and parameter estimates on the building characteristics in a similar model.

Third, we use demographic data about the neighborhoods from the 1980 U.S. Census to compare the characteristics of neighborhoods in which gardens were opened to those of other neighborhoods. Because most of the gardens were opened since 1980, the 1980 Census data seemed most appropriate.

Finally, we utilize data on the location and characteristics of most federally- and city-assisted housing in New York City from a variety of sources.³⁸ We obtained address-specific data from HUD USER on the number of units created through the Section 8 project-based, Section 202, LIHTC, BMIR and the Section 236 programs. From the New York City Housing Authority (NYCHA), we secured address-specific data on all public housing developments. New York City's Department of Housing Preservation and Development (HPD) provided address-specific data describing all of the city-assisted housing built between 1977 and 2000.

As mentioned above, identifying properties in the vicinity of garden sites was critical to our analyses. We used GIS techniques to measure the distance between garden sites and each property for which a sale appeared in our database. From these distance measures, we created a variable that identified properties within 1,000 feet of a garden. A continuous distance variable indicates the distance from the property sold to the closest garden site.³⁹

Table 1 shows summary statistics for our sales sample. The first column shows the characteristics of our full sample of property sales; the second column shows the characteristics of transacting properties that were located or in the future would be located within 1,000 feet of a garden. Fifteen percent of the sales were located within 1,000 feet of a garden site. As shown, most of the sales in our sample were located in Brooklyn and Queens, largely because those boroughs include a relatively large share of smaller properties, which sell more frequently than larger ones. Over two-thirds of all buildings sold were either one- or two-family homes, and 87% were single-family homes, two-family homes or small apartments.

³⁸ For more details on the subsidized housing data sets used in this article, see Ellen *et al.* (2007).

³⁹ As all tax lots in New York City have been geocoded by the New York City Department of City Planning we used a "cross-walk" (the "Geosupport File") which associates each tax lot with an *x, y* coordinate (*i.e.*, latitude, longitude using the U.S. State Plane 1927 projection), community district and census tract. In the case of physical structures, a tax lot is usually a building and is an identifier available to the property sales and RPAD data. We are able to assign *x, y* coordinates and other geographic variables to over 98% of the sales using this method. The community gardens data also include the tax lot(s) occupied by each garden, so we were able to assign *x, y* coordinates to each garden lot by the same method. If a garden encompassed multiple lots, we calculated the coordinates of the center of the garden.

Table 1 ■ Characteristics of residential properties sold.

	Percentage of All Property Sales	Percentage of Sales Within 1,000 Feet of Community Garden Sites
<i>Borough</i>		
Manhattan	8.0	17.2
Bronx	12.0	14.3
Brooklyn	37.2	53.3
Queens	37.4	14.6
Staten island	5.4	0.7
<i>Building class</i>		
Single-family detached	21.8	10.0
Single-family attached	13.9	9.0
Two-family	33.1	31.6
Walk-up apartments	18.1	31.5
Elevator apartments	1.0	2.0
Loft buildings	0.0	0.0
Condominiums	7.7	9.2
Mixed-use, primarily residential (Includes store or office plus residential units)	4.3	6.7
<i>Other structural characteristics</i>		
Built pre-World War II	78.4	89.4
Vandalized	0.0	0.2
Other abandoned	0.1	0.3
Garage	33.9	16.3
Corner location	7.3	7.5
Major alteration prior to sale	1.1	3.4
<i>n</i>	517,791	77,642

Note: Universe = all sales in community districts with community gardens.

The second column reveals some systematic differences between the transacting properties located close to garden sites and those that are not. Properties located within the 1,000-foot ring were more likely to be in Brooklyn and Manhattan than in the other boroughs. They were also older, less likely to be single-family homes and more likely to be walk-up apartments.

In an effort to better describe the small neighborhoods that host community gardens, we replicated the above comparisons using all properties in rings, not just those that sold. The same picture emerged using these statistics (available upon request from the authors). We also present in Table 2 statistics on the presence of publicly subsidized housing in the vicinity of garden sites. This table shows that the share of each type of assisted housing—except Mitchell Lama—in the total housing stock in the rings around garden sites is much larger

Table 2 ■ Subsidized housing within 1,000 feet of community gardens.

Program Type	Units			
	Within 1,000 Feet of Garden		Total	
	Number	% of All Residential	Number	% of All Residential
Public housing	90,189	14.1	169,569	7.1
Other federally subsidized ^a	50,257	7.9	90,068	3.7
City-subsidized homeownership	24,699	3.9	38,413	1.6
City-subsidized rental	93,914	14.7	163,746	6.8
Mitchell lama rental	15,710	2.5	53,576	2.2
Mitchell lama co-op	18,208	2.8	60,378	2.5
All residential	639,947	100.0	2,402,697	100.0

^aIncludes Sec. 8 NC/SR, Sec. 202, Sec. 236, BMIR, LIHTC.

Note: Universe = housing units in community districts with community gardens.

Table 3A ■ Descriptive statistics on New York City gardens.

<i>Garden Area (×1,000 sq. ft.)</i>	
Mean	35.5
Median	6.0
<i>Completion year</i>	
% established between 1977 and 1980	8.8
% established between 1981 and 1990	29.3
% established between 1991 and 2000	61.8
<i>Land ownership</i>	
% public	95.3
% private	4.7
<i>% Borough</i>	
Manhattan	27.2
Bronx	23.1
Brooklyn	43.1
Queens	6.1
Staten island	0.5
<i>Number of gardens</i>	636

than the share outside the rings. For example, 14.1% of the housing stock in rings consists of public housing units, but only 7.1% of the total city housing stock is public housing. Subsidized housing in the aggregate accounts for 46% of the housing stock in rings but only for 24% of the total stock.

Table 3 shows descriptive statistics for the sample including all of New York City gardens formed between 1977 and 2000, as well as for the subsample

Table 3B ■ Descriptive statistics on surveyed gardens in the Bronx.

<i>Garden Area (× 1,000 sq. ft.)</i>	
Mean	9.7
Median	6.1
<i>Completion year</i>	
% established between 1977 and 1980	8.2
% established between 1981 and 1990	36.1
% established between 1991 and 2000	55.8
<i>Land ownership</i>	
% public	95.3
% private	4.7
<i>Overall quality of garden</i>	
% gardens with acceptable overall quality ^a	76.7
% gardens with acceptable overall quality ^b	60.5
<i>Detailed garden quality</i>	
% gardens with acceptable community access	67.4
% gardens with acceptable fencing/security	64.0
% gardens with acceptable cleanliness	55.8
% gardens with acceptable landscaping	93.0
% gardens with (non-seasonal) decorations	53.5
% gardens with social spaces	34.9
<i>Number of gardens</i>	86

^aBased on surveyor's assessment.

^bBased on answers to all questions.

of Bronx gardens covered by our survey. Whereas the mean area of the city's gardens is 35,000 sq. ft., the median area is only 6,000 sq. ft., suggesting that the area distribution is highly skewed to the right. Most of the gardens are relatively new—62% were established within the last decade—and the overwhelming majority (95%) is sited on publicly owned land. Most of the gardens are located in Brooklyn, Manhattan and the Bronx, with Brooklyn exhibiting the highest concentration (43%). The gardens we surveyed regarding quality are remarkably similar to the whole New York City sample in terms of median area, completion year and land ownership.

Table 4 compares the average 1980 characteristics of census tracts that include gardens to those that do not have a garden but are in a subborough area that does have at least one garden.⁴⁰ It shows that gardens were generally located in

⁴⁰ We only use subboroughs with gardens for these statistics because our subsequent estimation of the garden impacts on property values is based only on these areas (recall

Table 4 ■ The 1980 characteristics of census tracts with and without gardens.

	Tracts With Gardens	Tracts Without Gardens But in SBA With Gardens
Mean family income	29,649	45,593
Mean poverty rate	36.7	18.8
Mean unemployment rate	13.2	8.1
Mean homeownership rate	12.7	31.5
Mean vacancy rate	5.8	2.9
Mean percentage of 25+yr old residents with some college education	16.8	25.2
Mean percentage black	49.8	25.1
Mean percentage Hispanic	31.8	18.8
Mean percentage foreign born	17.9	24.0
Mean percentage kids (age < 5 years)	10.0	7.8
Mean percentage age 5–17	23.2	18.4
Mean percentage age 18–64	57.1	59.9
Mean percentage old (age 65+)	9.7	13.9
Mean percentage of 5+yr old people who did not change address within last 5 years	57.3	61.0
N tracts	308	1316

Note: Tracts with less than 200 persons are excluded from the samples on which these statistics are based.

distressed neighborhoods.^{41,42} As compared to the average census tract without gardens, tracts with gardens had much lower mean family incomes, much higher poverty rates (twice as high) and unemployment rates, lower educational attainment, much lower homeownership rates (2.5 times lower) and higher vacancy rates. The tracts with gardens housed much greater shares of Hispanic and Black residents than the average tract without gardens. Finally, other demographic statistics indicate that the garden neighborhoods had smaller shares of

from footnote 35 that we limited the analysis to properties that are located within the community districts with community gardens; subborough area boundaries are, in general, fairly similar to community district boundaries, and, in addition, they match exactly with census tract boundaries).

⁴¹ The census tract data is taken from the 1980 Census. Tracts are characterized as including gardens even if these gardens did not open until later in the decade.

⁴² We use 1980 tract characteristics because Table 2 shows that a vast majority of the gardens in our study were built during the last two decades. Thus, the table largely captures characteristics of the tracts before the gardens were opened.

foreign-born population, a younger population and smaller shares of residents with stable neighborhood tenure.

Results

Baseline Model

Table 5 shows the key coefficients and their standard errors for our baseline model in Equation (1). Coefficients for structural variables are shown in Table A1 in the Appendix. The relatively high R^2 (0.86), together with the fact that the coefficients on the structural variables are consistent with expectations,

Table 5 ■ Selected regression results—baseline model.

<i>Ring Variables</i>	
InRing	−0.1109*** (0.0057)
InRing * D	9.9E-05*** (7.4E-06)
PostRing	0.0356*** (0.0109)
PostRing * D	−2.0E-05 (1.3E-05)
PostRing_GArea	−2.0E-04*** (5.8E-05)
PostRing_GArea * D	9.0E-08 (7.2E-08)
PostRing_GArea ²	6.4E-08* (3.6E-08)
PostRing_SharePvtOwn	−0.0282*** (0.0091)
Tpost	0.0062*** (0.0023)
Tpost * D	−7.3E-06*** (1.0E-06)
Tpost ²	4.5E-04** (2.2E-04)
Tpost ³	−1.6E-05** (6.2E-06)
<i>n</i>	517,791
Mean square error	0.0985
R^2	0.8560

Notes: Regression includes census tract and community district-quarter dummies and the full set of building controls, as in the appendix. Standard errors in parentheses. *** denotes 1% significance level; ** denotes 5% significance level; * denotes 10% significance level.

suggest that these variables provide adequate controls for the characteristics of the properties sold.⁴³

Looking at the coefficient estimates in Table 5, the first thing to notice is that the coefficient for the *InRing* variable (coded “1” if the property is within 1,000 feet of the site of an existing or future garden) is negative and statistically significant. In particular, prior to the date the garden was opened, properties located right next to a garden site ($D = 0$) sold for 11.1% less than comparable properties located outside the 1,000-foot ring. Anecdotal evidence suggests that sites on which gardens were built often were vacant, rubble-strewn sites that were havens for crime and other disamenities.⁴⁴ The negative effect the lots had on neighboring property values prior to the lots’ development as gardens is consistent with that evidence.⁴⁵ Although we can only speculate, another possible explanation for the negative *InRing* coefficient is that the lack of gardens in our control areas may reflect scarcity of land supply and thus (other things equal) tighter housing markets and higher property values in these areas, relative to the rings around gardens.⁴⁶

The coefficient for the *PostRing* variable (coded “1” if the property sold was within 1,000 feet of a garden) is positive and statistically significant. Its magnitude is 3.6 percentage points, indicating that the gap between prices in the ring and those outside the ring but in the same census tract narrowed after the garden was established.⁴⁷ Before the garden opened, the gap between properties

⁴³ Briefly, results indicate that sales price is higher if a building is larger or newer, located on a corner, or includes a garage. Sales price is lower if the building is vandalized or abandoned. The building class dummies are also consistent with expectations. Sales prices per unit for most of the building types are lower than those for single-family attached homes (the omitted category), and sales prices for single-family detached homes are higher than those for single-family attached homes. The coefficient on the dummy variable indicating that the building has undergone a major alteration prior to sale is positive and significant. Statistically significant coefficients on dummy variables indicating missing values for the age or size of a building indicate that these buildings missing age data are less valuable than others (perhaps because they are older) and buildings missing square footage data are more valuable (perhaps because they are larger).

⁴⁴ These derelict vacant lots were the result of urban decay, landlord disinvestment and abandonment during the fiscal crisis of the 1970s. Most of them were owned by the City and were temporarily leased to community gardeners for a symbolic fee.

⁴⁵ The coefficient for the *InRing * D* variable is positive and significant, indicating a sharp price gradient such that the pregarden price-depressing effects of the site (the disamenity) decline with distance. For example, at a distance of 1,000 feet, residential prices are only 1% lower, meaning that the price differential falls at a rate of about one percentage point per 100 feet.

⁴⁶ We thank an anonymous referee for suggesting this additional explanation.

⁴⁷ As noted before, the *PostRing* coefficient provides an estimate of the fixed component of the public garden effect—that is, the portion of the impact that is independent of the

adjacent to the garden site and out-of-ring properties was 11.1%; after the garden opened, this gap fell by 3.6 percentage points, to 7.5%.

$Tpost$ measures the number of years between the opening of the garden and the property sale. The coefficients on $TPost$ and its square are positive and significant, whereas the coefficient on $Tpost^3$ is negative and significant. This implies that the positive impacts gardens have on surrounding properties grow over time at an (slightly) increasing rate up to some point, beyond which the growth continues at a decreasing rate. The garden's increasing value to the neighborhood may result from greater certainty about the garden's maintenance, or perhaps may reflect neighbors' increasing awareness of, or involvement in, the garden. It also is consistent with the possibility that the gardens serve as catalysts of community development (see footnote 28 above). In the first year after the garden opens, the positive impact increases by 0.6 percentage points, and subsequent years bring additional and increasingly larger positive impacts. Ten years after opening, the growth starts slowing down, and it appears to stop 15 years later (*i.e.*, 25 years after opening). The negative and significant coefficient on the interaction between $Tpost$ and distance suggests that impacts increase less over time farther away from the garden.

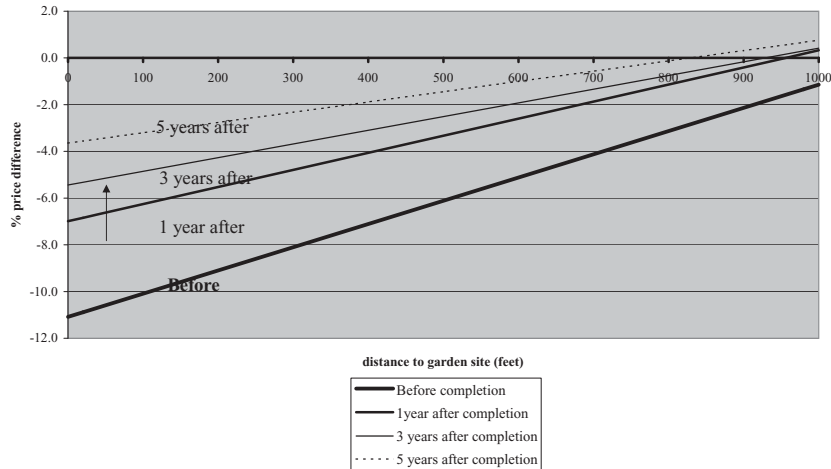
Figure 1 and Table 6 show estimated impacts for a typical New York garden with an area of 6,000 sq. ft.—the median garden size in our sample.⁴⁸ The thick line in Figure 1 shows the percentage difference between prices in the ring and prices in the surrounding neighborhood, before the garden was established. As noted, this pregarden gradient is fairly steep, climbing at a rate of 1 percentage point per 100 feet, such that sales prices for property nearest to the site are 11.1% lower than prices for properties outside the 1,000-foot ring.

The thinner lines above that baseline show price gradients one, three and five years after garden opening and suggest substantial impacts, especially for the

garden area. Increasing garden size appears to bring a smaller benefit, perhaps because larger gardens tend to be noisier and to generate more congestion. However, this negative marginal effect of garden area is fairly small and declines as the area increases. To take a concrete example, a 10,000 sq. ft. increase in garden area, from 1,000 to 11,000 sq. ft., reduces the external benefit by only 0.2 percentage points. Spillover benefits are also negatively affected by the private ownership of the land on which the garden is sited. The benefit of a privately owned garden is 2.8 percentage points lower than that of similar garden sited on publicly owned land, and it is not statistically significant. Finally, note that the coefficients on $PostRing * D$ and $PostRing_GArea * D$ are not statistically significant, suggesting that impacts vary little with distance from the site right after garden opening.

⁴⁸ We use the median rather than the mean garden size to define the typical garden because our sample distribution of garden area is highly skewed to the right (*e.g.*, the mean is larger than the 80th percentile).

Figure 1 ■ Percentage difference between residential prices in 1,000-foot ring and surrounding neighborhood, before and after completion of a typical garden, by distance to garden site and time since completion.



Note: The typical garden is considered one which is publicly-owned and has an area of 6,000 sq.ft. (the median garden size in our sample).

housing closest to the garden. Before the garden opens, prices in the immediate vicinity of the garden site are 11.1% lower than in the surrounding neighborhood; in the first year after opening this gap falls by 4.1 percentage points, to 7%.⁴⁹ Over the five-year period after opening, the gap falls by 7.4 percentage points, to only 3.7%. Note that at 1,000 feet, impacts are smaller, ranging from 1.5 percentage points one year after the garden opening to 1.9 percentage points five years after opening.⁵⁰

Table 6 also includes the dollar-value of garden impacts, estimated by applying the percentage-point impacts to the median real sale price—expressed in 2003 dollars—for housing in the 1,000-foot rings around gardens (\$88,032).⁵¹ For

⁴⁹ The difference between the 3.6 percentage point impact reported in Table 5 and the 4.1 percentage point impact reported in this simulation results partly from our using a garden of the median size in our sample for the simulation. An additional reason for this discrepancy is that the former number represents the impact immediately after garden opening whereas the latter represents the impact one year after opening.

⁵⁰ Note, however, that these estimates likely understate impacts at 1,000 feet because the coefficients on the interactions of *PostRing* and *GardenArea* with distance are not statistically significant, and yet we included them in predictions. Discarding the insignificant coefficients yields impact estimates at 1,000 feet that vary between 3.3 percentage points one year after opening to 3.8 percentage points five years after opening.

⁵¹ We use median rather than mean price in our simulations because the citywide mean house price is driven up by the hot submarkets in Manhattan and certain areas of

Table 6 ■ Impacts of a typical garden on residential property values (estimates based on median per-unit sales price of properties in rings).

Time Since Completion	Distance to Garden Site (feet)					
	0		500		1000	
	%	\$-value	%	\$-value	%	\$-value
Right after completion	3.4	3,027	2.5	2,191	1.5	1,355
1 year	4.1	3,607	2.8	2,450	1.5	1,293
3 years	5.6	4,971	3.6	3,172	1.6	1,373
5 years	7.4	6,551	4.7	4,111	1.9	1,670

Notes: % impact is percentage point change in gap between prices in ring and outside; \$-value impact is obtained by applying the % impact to the median per unit sales price of properties sold in rings. Median price per unit of properties sold in rings is \$88,032. The typical garden is considered one that is publicly owned and has an area of 6,000 sq.ft. (the median garden size in our sample).

example, in the immediate vicinity of the garden, the dollar-value impact is \$3,607 one year after opening, and grows to \$6,551 five years after opening.

In summary, we find that gardens were located on sites that acted as local disamenities within their communities. After opening, gardens have a positive impact on surrounding property values, which grows steadily over time, and declines somewhat with garden area and distance to the garden.⁵²

Heterogeneity of Impacts Across Different Neighborhood Types

Table 7 reports the results of the model in which impacts are allowed to vary between lower and higher-income submarkets. Coefficients in column 1 capture impacts in higher-income submarkets; coefficients in column 2 capture the difference between the impacts in low- and higher-income submarkets.

The estimates reveal significant differences in garden impacts between the low- and higher-income neighborhoods. The coefficient for the *PostRing* variable (indicating that the property is within 1,000 feet of a site on which a garden has been established) is not statistically significant in higher-income areas, but it is positive, significant and relatively large in low-income areas.⁵³ These

Brooklyn. Nonetheless, we also computed the less conservative dollar-value impact estimates based on the mean housing price in rings (\$125,275). Right next to the garden, these estimates range from \$5,134 one year after opening to \$9,323 five years after opening (more detailed estimates are available upon request from the authors).

⁵² Impact also declines with private ownership of the land on which the garden is sited.

⁵³ Additionally, in low-income neighborhoods, the negative marginal effect of additional garden area is significantly smaller than in higher-income neighborhoods.

Table 7 ■ Model with neighborhood heterogeneity.

	High Income Submarket (1)	Low Income – High Income Differential (2)
<i>Ring variables</i>		
InRing	–0.0469*** (0.0102)	–0.0930*** (0.0123)
InRing * D	3.9E-05*** (1.3E-05)	8.7E-05*** (1.6E-05)
PostRing	–0.0020 (0.0206)	0.0570** (0.0243)
PostRing * D	2.2E-05 (2.6E-05)	–6.4E-05** (3.0E-05)
PostRing_GArea	–3.6E-04*** (9.6E-05)	2.5E-04** (1.2E-04)
PostRing_GArea * D	6.8E-08 (1.0E-07)	4.1E-08 (1.4E-07)
PostRing_GArea ²	2.2E-07*** (5.8E-08)	–2.4E-07*** (7.4E-08)
PostRing_SharePvtOwn	–0.0411*** (0.0154)	0.0337* (0.0192)
Tpost	0.0039 (0.0043)	0.0032 (0.0051)
Tpost * D	–6.9E-06*** (2.1E-06)	4.6E-07 (2.4E-06)
Tpost ²	8.0E-04** (4.1E-04)	–5.9E-04 (4.8E-04)
Tpost ³	–3.1E-05*** (1.2E-05)	2.5E-05* (1.4E-05)
<i>n</i>		517791
Mean square error		0.0974
<i>R</i> ²		0.8576

Notes: Coefficients in column (2) correspond to a set of interactions between the ring variables and a dummy, which is equal to 1 for the low submarket and 0 otherwise. The low-income submarket comprises community districts for which the CD/MSA mean household income ratio is smaller than 0.8 (and the high income submarket includes all the other community districts). Regression includes census tract and CD-quarter dummies, and the full set of building controls and their interactions with the low-income submarket dummy. Standard errors in parentheses. *** denotes 1% significance level; ** denotes 5% significance level; * denotes 10% significance level.

differences suggest that, while there is little positive impact of gardens in higher-income neighborhoods (at least not soon after opening) gardens are likely to generate significant and sustained benefits in poorer neighborhoods.⁵⁴ We estimate

⁵⁴ Note, however, that even in higher-income areas benefits may arise over time, as

that the price differential between a property located right next to a typical garden and a comparable property located outside the ring falls by 6.2 percentage points one year after a garden opens in low-income submarkets. This translates into a dollar-value of \$4,618, based on the median real sale price for housing within 1,000 feet of gardens in low-income areas (\$75,000). Five years after the garden opens, the impact is even larger: 9.4 percentage points or \$7,071. Again, impacts are smaller farther away from the garden: at 1,000 feet they range from 1.3 percentage points one year after opening to 2.1 percentage points five years after opening (more detailed estimates are available upon request from the authors).

The finding that community gardens bring significantly larger benefits in lower-income neighborhoods is not surprising. As pointed out in several prior studies (Francis, Cashdan and Paxson 1984, Harnick 2000, Saldivar-Tanaka and Krasny 2004), gardens in poor neighborhoods provide an affordable alternative to city parks, which often are located in more affluent neighborhoods and are not easily accessible to poor residents. Moreover, the residents of lower-income neighborhoods are predominantly renters who, for reasons noted earlier, may be more interested in gardens than homeowners.⁵⁵

Given that subsidized housing does not sell often or at all (like public housing), that it makes up a significant share of the housing stock in the rings around gardens (as shown in Table 2), and that it is heavily concentrated in the poor neighborhoods where the gardens bring larger benefits, our average impact estimates for the city as a whole (from Table 5) may be downward biased.⁵⁶

Garden Quality

As mentioned in the methodology section, we were unable to rate the quality of all the gardens in our sample, so we focused our quality survey on gardens in the Bronx. Our analysis of whether and how quality matters, therefore, must use estimates of our baseline specification based on the surveyed Bronx gardens only. These estimates are presented in Table 8. The results are in line with our general findings—gardens have a statistically significant positive effect on in-ring properties. The results also are consistent with our analysis of the

suggested by the positive and significant $TPost^2$ coefficient.

⁵⁵ We alternatively hypothesized at the outset that homeowners may have stronger economic incentives than renters to create higher-quality gardens, but we did not find a statistically significant correlation between garden quality and the homeownership rate in the garden neighborhood.

⁵⁶ See Schill *et al.* (2002) and Ellen *et al.* (2006) for comprehensive evidence that subsidized housing is concentrated in poor neighborhoods.

Table 8 ■ Baseline model based on survey sample.

<i>Ring Variables</i>	
InRing	-0.1219*** (0.0178)
InRing * D	1.0E-04*** (2.3E-05)
PostRing	0.1016*** (0.0292)
PostRing * D	-1.1E-04*** (3.6E-05)
PostRing_GArea	-0.0021** (8.9E-04)
PostRing_GArea * D	-7.1E-07 (8.6E-07)
PostRing_GArea ²	1.6E-05*** (5.5E-06)
PostRing_SharePvtOwn	-0.0501 (0.0000)
Tpost	0.0125** (0.0062)
Tpost * D	1.6E-06 (2.8E-06)
Tpost ²	-0.0012** (6.0E-04)
Tpost ³	3.6E-05** (1.7E-05)
<i>n</i>	61,880
Mean square error	0.0691
<i>R</i> ²	0.8756

Notes: Regression includes census tract and community district-quarter dummies and the full set of building controls. Standard errors in parentheses. *** denotes 1% significance level; ** denotes 5% significance level; * denotes 10% significance level.

neighborhood heterogeneity of impacts; that is, the positive impacts of gardens are larger in lower income areas, such as the Bronx. Note, however, that in the Bronx model, the positive *PostRing* coefficient and the negative coefficients of *PostRing * D* and garden area are substantially larger than the low-income submarket coefficients estimated for the whole city. These suggest that the positive impacts of Bronx gardens, although very large for small gardens and in the immediate garden proximity, decline relatively quickly with garden area and distance to the garden.

The garden quality measures summarized in Table 3 show that between 60% and 77% of the surveyed gardens (depending on the assessment method) have acceptable overall condition. About two-thirds of the surveyed gardens have acceptable community access and fencing quality/security; more than half are

acceptable in terms of cleanliness and have some sort of (nonseasonal) decorations; a large majority (93%) have acceptable landscaping; and relatively few gardens (35%) have some kind of social gathering spaces.

We next turn to the results of an extension of the baseline model that includes variables describing the overall quality of gardens.⁵⁷ Table 9 shows the estimates of a specification that includes the “Acceptable/Not Acceptable” measure of the overall garden condition. The negative and significant coefficient on the *InRing_ShareAQ* variable indicates that neighborhoods with a larger share of garden area of acceptable quality were, prior to the garden openings, relatively more distressed than neighborhoods with a smaller share of garden area of acceptable quality. For example, prior to the garden openings, the (average) relative price⁵⁸ of properties located within 1,000 feet of sites that would host good-quality gardens only (*i.e.*, properties in ring for which *InRing_ShareAQ* = 1) was eight percentage points lower than the relative price of properties located within 1,000 feet of sites that will ultimately hold low-quality gardens only (*i.e.*, properties in ring for which *InRing_ShareAQ* = 0).⁵⁹

Turning to impacts, the *PostRing_ShareAQ* coefficient indicates that impacts increase significantly with the share of existing garden area with acceptable quality within 1,000 feet of the property; an increase in this share from 0 to 1 is associated with a 7.8 percentage point increase in impacts.

Again, simulations are helpful to summarize results. Assuming, for the sake of a more intuitive example, that properties are within 1,000 feet of only one garden,⁶⁰ the impact of a low-quality garden of typical size on the properties located in its immediate vicinity ($D = 0$) is five percentage points and is statistically significant at the 10% level, whereas that of a good-quality garden of similar size is 12.8 percentage points and highly significant.⁶¹ As distance to the garden increases, however, the positive impact of low-quality gardens vanishes (at around 440 feet) and is transformed into an increasingly negative effect. The

⁵⁷ Table A2 in the Appendix presents results of a specification that focuses on specific aspects of garden quality identified in the survey—community access, fencing attractiveness/permanence, cleanliness, landscaping, decorations and social spaces. We find that features that matter most are fencing attractiveness/permanence, cleanliness and landscaping.

⁵⁸ By relative price, we mean the percentage price differential between properties in the ring and those outside the ring (but still in the same census tract).

⁵⁹ In this discussion, we are using interchangeably the terms “acceptable (unacceptable) condition” and “good (low)-quality.”

⁶⁰ This assumption is also realistic, given that a majority of property sales are within 1,000 feet of only one garden (see footnote 14).

⁶¹ These estimates are computed at one year after garden opening.

Table 9 ■ Model with overall quality of garden.

<i>Ring Variables</i>	
In Ring	-0.0653*** (0.0212)
In Ring * D	9.5E-05*** (2.3E-05)
<i>Overall quality of future or existing gardens</i>	
InRing_ShareAQ	-0.0797*** (0.0160)
Post Ring	0.0500 (0.0313)
Post Ring * D	-1.1E-04*** (3.6E-05)
PostRing_GArea	-2.0E-03** (8.9E-04)
PostRing_GArea * D	-7.4E-07 (8.6E-07)
PostRing_GArea ²	1.6E-05*** (5.5E-06)
PostRing_SharePvtOwn	-0.0393 (0.0333)
Tpost	0.0129** (0.0062)
Tpost * D	1.9E-06 (2.8E-06)
Tpost ²	-1.3E-03** (6.0E-04)
Tpost ³	3.9E-05** (1.7E-05)
<i>Overall quality of existing gardens at the time of sale</i>	
PostRing_ShareAQ	0.0780*** (0.0167)
<i>n</i>	61,880
Mean square error	0.0691
<i>R</i> ²	0.8756

Notes: All regressions include census tract and community district-quarter dummies and the full set of building controls. Standard errors in parentheses. *** denotes 1% significance level; ** denotes 5% significance level; * denotes 10% significance level.

positive impact of good-quality gardens, diminishes, too, with distance, and at 1,000 feet it is statistically insignificant.

These results are robust across specifications using alternative measures of the overall garden condition (described in the methodology section above). The estimates of these alternative specifications are available upon request from the authors.

Alternative Specifications

As noted above, the average price differential between the rings and their census tracts may have already been declining prior to the garden establishment and may have continued to decline, even without the garden. To test for this possibility, we extend our specification to include the trend in housing prices before the garden opens within the ring surrounding what will become a garden. We do find a slight upward trend in residential prices prior to the garden opening, however, this trend is not statistically significant.⁶² Moreover, even after controlling for these prior trends in housing prices, we still find significant positive impacts upon garden opening.⁶³ In fact, the impact estimates are only slightly smaller than those found in the baseline model. The positive impacts gardens have on residential properties accordingly are not explained by a general upward trend in prices in the ring.

To account for the possibility that publicly subsidized housing investments may have systematically occurred at the same times and locations as the community gardens, thus biasing our garden impact estimates, we also estimated models in which we include extensive controls for proximity to various types of subsidized housing. The results remained virtually unchanged, indicating that this scenario is unlikely.⁶⁴

Finally, we estimate a repeat sales specification to test for the possibility that our results simply capture the fact that higher-quality properties were sold after the garden openings. We obtain repeat sales estimates that are similar to the hedonic ones; if anything, they yield somewhat larger spillover benefits than the hedonic estimates.⁶⁵

Net Tax Benefits of Community Gardens

Some jurisdictions are considering using either tax increment financing or impact fees to finance the provision of parks and small green spaces. To help inform those discussions, we provide here an estimate of the net tax benefits community gardens have brought to New York City.

The City usually has incurred little or no expense, other than opportunity costs, related to the development and maintenance of the existing community gardens. These costs usually were supported by funds local community groups raised

⁶² We also find a significant increase in house prices in the year immediately preceding the garden opening, however, as noted above, this increase is likely caused by the garden itself.

⁶³ These estimates are available upon request from the authors.

⁶⁴ These estimates are available upon request from the authors.

⁶⁵ The repeat sales estimates are available upon request from the authors.

from local sources or from foundations, private individuals or federal or state government agencies. However, from a policy perspective, it would be informative to consider alternative scenarios in which some or all costs are paid for through public subsidies.

To quantify garden costs, we rely on estimates provided in an earlier study by Fox, Koeppel and Kellam (1985), which contains perhaps the most comprehensive cost analysis of New York City community gardens. Fox, Koeppel and Kellam (1985), surveyed a large group of community gardens in New York City (over 300 gardens), then estimated that the initial capital costs for an average garden,⁶⁶ including all one-time expenses⁶⁷ applied to the original development of the site, would total \$16,365.⁶⁸ They also estimated an annual maintenance cost (including sweat equity and expenditures on items needed in the garden maintenance) of \$8,683.⁶⁹ To obtain the total garden cost, we add the net present value of the stream of maintenance costs over a 20-year period to the (one-time) initial capital costs.⁷⁰ In the end, we estimate that a garden would cost, on average, about \$279,000 (in 2003 dollars) to develop and maintain for 20 years.

To simulate the gross tax revenue gains generated by community gardens, we estimate the aggregate increase in residential property values generated by each garden (within a 1,000-foot radius) and then apply a standard tax assessment formula to these benefits, following the approach in Schwartz *et al.* (2006). Specifically, to estimate the aggregate benefits, we undertake a three-step process. First, we use the RPAD database⁷¹ to identify all the residential properties (whether sold or not) that were located within 1,000 feet of a garden site and thus should have benefited from the garden establishment.

Second, we assign an approximate initial or “preopening” price to each of these properties. We calculated, for each year in our data, the median per unit sales price for all properties that sold in that year within 1,000 feet of any garden

⁶⁶ The average garden size in Fox, Koeppel and Kellam (1985) is 5,225 sq. ft., which is very similar to the median garden size in our study (6,000 sq. ft.)

⁶⁷ These initial, one-time expenses cover the clearing of the site, providing a fence, plant materials, lumber of raised beds and benches, signs, technical assistance and the like.

⁶⁸ This dollar value represents 1982 dollars. In 2003 dollars, the cost is \$31,500.

⁶⁹ Again, this dollar value represents 1982 dollars. In 2003 dollars, the annual maintenance cost is \$16,700.

⁷⁰ For the present value calculation, we assume a growth rate in maintenance costs of 4.2% (equal to the average inflation rate over the study period) and a discount rate of 5.35% (the average for recent city general obligation debt issues). We compute maintenance costs over a 20-year period because this is the time interval used to estimate tax benefits (see below).

⁷¹ The RPAD database is an annual census of all New York City properties, described above.

site, prior to the establishment of the garden. The “preopening price” assigned to each property (sold or not) is equal to the median price that corresponds to the establishment year of the garden within 1,000 feet of that property.

Third, we use the preopening prices and number of units of each property, the size and land ownership of gardens within 1,000 feet of each property, the distance from each property to the nearest garden site and our baseline postopening coefficients, to estimate the increase in property values that should have occurred following garden completion. The sum of these gains over all properties is an estimate of the total benefit that all the gardens delivered to residential properties within a 1,000-foot radius. To be conservative, we assume that the growth in impacts over time (given by the *TPost* coefficient) ends five years after garden establishment. Given this assumption, we compute total benefits as the net present value of the five annual increments.⁷²

Using this three-step method, we estimate that the benefits generated in the 1,000-foot ring total almost \$1.3 billion (in 2003 dollars)—or \$2 million per garden.⁷³ To measure the corresponding tax revenue gains, we first discount total benefits by a factor of 0.7 as the appraised market values used by the New York City Department of Finance for tax assessment purposes are, on average, about 70% below sales prices (on which our measured benefits are based). We then estimate the increase in tax revenues in the first year after garden completion by applying the New York City assessment ratios and tax rates to the discounted increases in property values accrued during that year.⁷⁴ Finally, we compute the total tax revenue gains as the present value of the stream of annual tax benefits over a 20-year period, based on assumptions suggested by the staff of the New York City Independent Budget Office.⁷⁵

⁷² For this calculation, we assume, again, a temporal discount rate of 5.35%.

⁷³ A comparison between the 1998 assessed values of properties in rings that sold in 1999 and the assessed values of all properties in rings suggests that transacting properties may have higher values than other properties. We thus correct for this selection bias when estimating total benefits.

⁷⁴ Assessment ratios and tax rates vary by property type. The Department of Finance groups residential properties into three classes—class 1 (1–3 family houses), class 2 (4–6 family buildings) and class 2A (all the others)—and sets an assessment ratio and a tax rate for each class. The assessment ratio for class 1 is 0.08, for class 2 is 0.25 and for class 2A is 0.45. The tax rate for class 1 is 0.116 and for the other two classes is 0.108. Due to this variation in assessment and tax rates, we also estimate the total benefits from gardens separately for each of the three building classes and then apply the corresponding rates to the benefit for each class.

⁷⁵ Specifically, we assume an annual growth of 2.5% in property taxes over 20 years, based on average increases in assessments and a discount rate of 5.35%. Note that during the first five years after garden completion, the assumed 2.5% annual increase in taxes is over and above the increase due to the growth in garden impacts.

In the end, we estimate that the present value of the gross tax benefit to the City generated by all community gardens over a 20-year period amounts to about \$503 million. If the City had fully subsidized the garden provision, the City's total investment would have amounted to over \$177 million. Thus, the estimated present value of the net tax benefit over 20 years would be, in the aggregate, over \$325 million or, per garden, about \$512,000.

Of course, in cities facing shortages of developable land, the major cost of community gardens will be the opportunity costs of not developing the land for other uses. It may be, for example, that using the lots to develop housing would boost neighboring property values, and the jurisdiction's consequent tax receipts, by more than the benefits associated with community gardens. Estimating the opportunity costs of the community gardens is beyond the scope of this article.

Conclusion

We find that community gardens have, on average, significant positive effects on surrounding property values, and that those effects are driven by the poorest of host neighborhoods (where a garden raises neighboring property values by as much as 9.4 percentage points within five years of the garden's opening). Those findings should help local governments make sounder decisions about whether (and how much) to invest in (or to encourage private investment in) community gardens and other green spaces. Such investments have a sizeable payoff for the surrounding community, and ultimately for the city itself, as it realizes additional property tax revenues from the neighborhood.

Our findings also will help local governments considering whether to use tax increment financing (TIF) to estimate the potential benefits of investments in urban parks and gardens. Our results show that such gardens can lead to increases in tax revenues of about half a million dollars per garden over a 20-year period. Finally, local governments may use our results to justify the imposition of impact fees to finance the provision of gardens or urban parks, by showing the benefits the developers' properties will receive as a result of proximity to such spaces.

Given that our study focuses exclusively on New York City, we are cautious in generalizing our findings and their implications to other settings, especially those outside of urban environments. That said, the great diversity of New York City's neighborhoods and gardens means that the estimated impacts reflect the impact of a variety of gardens in a range of neighborhoods, and that increases our confidence that similar results can be obtained in other cities. Moreover, preliminary results from our ongoing comparative analysis of the location of gardens and other open spaces suggest that garden impacts do not vary much

with the amount of open space available in the larger community, which again bolsters our view that these findings may be relevant to less dense settings.

We would like to thank Caroline Bhalla, XuFeng Chen and Drew Schinzel for their hard work and good spirits in securing and processing the data necessary for the project; Geoff Davenport and the team of students he lead to conduct the garden quality survey (Amanda Garcia, Ilmi Granoff, Keilem Ng, Gila Jones, Josh Frankel, Tammy Kim, Dave Gunton, Margaret Barry, Ansley Samson, Uma Deshmukh, Avery Wentzel, Matthew Johnston, Peter Hyun, Ed Kim, Marissa Soto and Katie Renshaw); and Caroline Bhalla, Ingrid Gould Ellen, Solomon Greene, Rachel Meltzer, Sessa Pochiraju, Richard Revesz, Drew Schinzel, Amy Ellen Schwartz and participants in the NYU Law and Economics Workshop, the NYU Colloquium on the Law, Economics and Politics of Urban Affairs, the University of Michigan Law School Law and Economics Workshop, the University of Connecticut Storrs Lecture Series, the Georgia State University 2006 Law Review Symposium, the Furman Center Round Table on Community Gardens and the 2006 AREUEA Mid-Year Meeting for their helpful critiques of the study. The financial support of the Max and Filomen D'Agostino Research Fund and the Furman Center for Real Estate and Urban Policy is gratefully acknowledged.

The views expressed in this article are those of the authors and do not necessarily reflect those of the Office of the Comptroller of the Currency or the Department of the Treasury. Ioan Voicu contributed to this article while a research scholar at the Furman Center for Real Estate and Urban Policy, New York University.

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Appendix
Table A1 ■ Complete regression results for baseline model.

<i>Ring Variables</i>	
In Ring	-0.1109*** (0.0057)
In Ring * D	9.9E-05*** (7.4E-06)
PostRing	0.0356*** (0.0109)
PostRing * D	-2.0E-05 (1.3E-05)
PostRing_GArea	-2.0E-04*** (5.8E-05)
PostRing_GArea * D	9.0E-08 (7.2E-08)
PostRing_GArea ²	6.4E-08* (3.6E-08)
PostRing_SharePvtOwn	-0.0282*** (0.0091)
Tpost	0.0062*** (0.0023)
Tpost * D	-7.3E-06*** (1.0E-06)
Tpost ²	4.5E-04** (2.2E-04)
Tpost ³	-1.6E-05** (6.2E-06)
<i>Ring variables for gardens with invalid completion year¹</i>	
In Ring	-0.0201* (0.0105)
Post Ring	0.0078 (0.0107)
PostRing_GArea	-3.4E-07*** (4.3E-08)
<i>Characteristics of properties sold</i>	
Vandalized	-0.0771*** (0.0243)
Other abandoned	-0.0868*** (0.0138)
Odd shape	0.0225*** (0.0017)
Garage	0.0491*** (0.0012)
Extension	0.0484*** (0.0016)
Corner	0.0493*** (0.0018)
Major alteration prior to sale	0.0454*** (0.0048)
Age of unit	-0.0103*** (0.0001)

Table A1 ■ continued

(Age of unit) ²	6.8E-05*** (8.4E-07)
Age of unit missing	-0.2148*** (0.0042)
Log square feet per unit	0.4120*** (0.0016)
Number of buildings on same lot	-0.0410*** (0.0027)
Includes commercial space	0.0284*** (0.0051)
Square feet missing	2.8896*** (0.0200)
Condo and square feet missing	-0.1123*** (0.0170)
Single-family detached	0.0881*** (0.0017)
Two-family home	-0.3354*** (0.0017)
Three-family home	-0.5773*** (0.0022)
Four-family home	-0.7579*** (0.0034)
Five-/six-family home	-1.1274*** (0.0039)
More than six families, no elevator	-1.5211*** (0.0040)
Walkup, units not specified	-1.3727*** (0.0054)
Elevator apartment building, cooperatives	-1.3830*** (0.0092)
Elevator apartment building, not cooperatives	-1.5101*** (0.0061)
Loft building	-0.7708*** (0.0244)
Condominium, single-family attached	-0.2395*** (0.0113)
Condominium, walk-up apartments	-0.2446*** (0.0054)
Condominium, elevator building	-0.5052*** (0.0050)
Condominium, miscellaneous	-0.7777*** (0.0102)
Multiuse, single family with store	-0.0865*** (0.0071)
Multiuse, two family with store	-0.5488*** (0.0062)

Table A1 ■ continued

Multiuse, three family with store	-0.7706*** (0.0086)
Multiuse, four or more family with store	-0.9609*** (0.0067)
<i>n</i>	517,791
Mean square error	0.0985
<i>R</i> ²	0.8560

Notes: Regression includes census tract and community district-quarter dummies. Standard errors in parentheses. *** denotes 1% significance level; ** denotes 5% significance level; * denotes 10% significance level. ¹Gardens with invalid completion year include gardens completed before 1977 or after 2000 and gardens with missing completion year. For the later, we assume that the completion year is prior to the earliest sale date in our sales sample, so that we can construct the post variable.

Table A2 ■ Model with specific quality features.

<i>Ring Variables</i>	
In Ring	-0.1014*** (0.0201)
In Ring * D	1.0E-04*** (2.3E-05)
<i>Overall quality of future or existing gardens</i>	
InRing_ShareAQ	-0.0351*** (0.0137)
Post Ring	0.0404 (0.0341)
Post Ring * D	-1.1E-04*** (3.7E-05)
PostRing_GArea	-2.4E-03*** (9.0E-04)
PostRing_GArea * D	-9.4E-07 (8.6E-07)
PostRing_GArea ²	1.6E-05*** (5.6E-06)
PostRing_SharePvtOwn	-0.0167 (0.0351)
Tpost	0.0110* (0.0062)
Tpost * D	1.2E-06 (2.8E-06)
Tpost ²	-1.1E-03* (6.0E-04)
Tpost ³	3.3E-05* (1.7E-05)

Table A2 ■ continued

<i>Detailed garden quality at the time of sale</i>	
PostRing_ShareACA (acceptable community access)	0.0078 (0.0196)
PostRing_ShareAF (acceptable fencing attractiveness and permanence)	0.0295* (0.0155)
PostRing_ShareAC (acceptable cleanliness)	0.0382** (0.0194)
PostRing_ShareAL (acceptable landscaping)	0.0575** (0.0251)
PostRing_SharePD (presence of decorations)	-0.0184 (0.0144)
PostRing_SharePS (presence of social spaces)	-0.0094 (0.0151)
<i>n</i>	61,880
Mean square error	0.0691
<i>R</i> ²	0.8756

Notes: Regression includes census tract and community district-quarter dummies and the full set of building controls. Standard errors in parentheses. *** denotes 1% significance level; ** denotes 5% significance level; * denotes 10% significance level.