

1 **The Closer the Better? Examining Support for a Large Urban** 2 **Redevelopment Project in Atlanta**

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11 12 **Abstract**

13 The Atlanta BeltLine (BeltLine) is a large urban redevelopment project that is
14 transforming 22 miles of historical railroad corridors into parks, trails,
15 pedestrian-friendly transit, and affordable housing in the center of Atlanta, Georgia.
16 This study examines how proximity to the BeltLine and other factors relate to public
17 support for it, with data from a general public survey conducted in the summer of
18 2009. The result shows that support significantly declines as distance to the
19 BeltLine increases. However, after controlling for expected use of the BeltLine parks
20 and transit, the role of distance fades. Further, the results show that being a parent
21 within the city limits is associated with the support for the BeltLine, which implies
22 that the concern over Tax Increment Financing (TIF) affecting future school quality
23 hampers the support of the project. The findings point to individual tastes and
24 family circumstances as driving support for the redevelopment project, rather than
25 strictly property-specific attributes (as the homevoter hypothesis would predict).

26 Another contribution of this study is to address the technical problem of
27 missing precise spatial location values. Several imputation techniques are used to
28 demonstrate the risks and remedies to missing spatial data.

29 30 **1. Introduction**

31 This article uses the Atlanta BeltLine (BeltLine) as a case study to examine
32 the relationship of distance and other factors to public support for urban
33 redevelopment projects. The BeltLine is a large, multibillion-dollar urban
34 redevelopment project centered in Atlanta, Georgia. The core purpose of the project
35 is to transform 22 miles of historic railroad corridors into pedestrian-friendly rail
36 transit, multi-use trails, parks, and affordable housing. The BeltLine is currently one
37 of the largest urban redevelopment and mobility projects in the United States and its

1 proponents have noted that it is able to “transform the city” of Atlanta (Atlanta
2 Development Authority, 2005; Kirkman, Noonan, & Dunn, 2012). Thus, interest in
3 and impacts of the BeltLine project are expected to extend broadly through the
4 Atlanta urban area. The Atlanta BeltLine project presents another example of major
5 urban redevelopment projects that seek to transform urban form, such as the High
6 Line in New York City (Loughran, 2014), Cheonggyecheon restoration in Seoul (Lee
7 & Anderson, 2013), and urban regeneration in Barcelona (Degen & GarcÍA, 2012).

8 The association between distance to the BeltLine and public support for it
9 could result from a variety of factors. First, being closer to amenities such as new
10 parks, trails, and public transit is expected to be positively related to residents’
11 support, especially for those who regularly use these amenities. Second, even those
12 who do not directly enjoy these amenities may still benefit from the increase of
13 property values because of the increase in amenities, as posited by the homevoter
14 hypothesis. First developed by Fischel (2005), the homevoter hypothesis holds that
15 homeowners politically support actions of local governments that increase their
16 property values. In this case, local homeowners are expected to support the BeltLine
17 as long as the project increases their property values. The property value increase
18 could be due to the increase of actual or anticipated accessibility to amenities or due
19 to the perception of “Atlanta being a better place.” Since being closer to the
20 BeltLine is expected to yield a higher property value premium, which makes
21 homeowners more supportive, the distance is theoretically negatively related to
22 support. This study tests the homevoter hypothesis by determining how public
23 support varies with distance, under the assumption that property value increases
24 because of the BeltLine are correlated with the distance to it.

1 On the other hand, support for the BeltLine could be hampered by projects
2 aiming to increase housing supply and affordable housing in particular. According
3 to Atlanta BeltLine, Inc. (ABI), the BeltLine project aims to create 28,000 new
4 housing units, 5,600 of which as affordable housing, over twenty-five years (Atlanta
5 BeltLine Inc., 2013). Under the homevoter hypothesis, homeowners near the
6 BeltLine’s affordable housing may oppose the project because a greater housing
7 supply would be harmful to their property values. However, while some previous
8 studies confirm the negative effect of affordable housing projects (Santiago, Galster,
9 & Tatian, 2001), the direction of the external neighborhood effect is still under debate
10 (Deng, 2011).

11 Further, a common redevelopment tool, Tax Increment Financing (TIF), may
12 be another important factor affecting support and the role of distance. TIF allows
13 local governments to fund particular projects with the future growth in property taxes
14 – the increment – created by the project itself. Mainly funded by TIF, the BeltLine
15 will essentially pay for itself by the property tax increment collected in the Tax
16 Allocation District (TAD) over the next 25 years. By reserving that increment for
17 servicing the debt incurred to implement the BeltLine, however, the TIF blocks the
18 use of future tax revenue growth for other categories, especially public education, for
19 a period of 25 years (Brueckner, 2001). In the U.S., local governments provide
20 nearly half of public school system revenue, and 66 percent of local revenue derives
21 from property taxes (McGuire, Papke, & Reschovsky, 2015). The reallocation of
22 future property tax revenue is expected to lower the quality of public school in the
23 future if educational costs keep growing. With the BeltLine-induced population
24 growth, it is thus expected that residents with children, especially those who plan to
25 send their kids to public schools, may favor the BeltLine less. Since the Atlanta

1 Public Schools district (APS) can redistribute budget shocks around its system, all
2 public schools in APS can be affected. Thus, households with children in the APS
3 jurisdiction may support the BeltLine less.

4 Another focus of this study is the problem of missing precise spatial location
5 information, a common challenge in empirical work. Data used in this study are from
6 a survey conducted in the summer of 2009 that asked about opinions and expectations
7 about the BeltLine. One limitation of using these survey data for this study is that
8 only half of the respondents provide their actual addresses (the rest gave only zip
9 codes). This information may not be missing at random. To solve this
10 missing-spatial-location problem, this study attempts several imputation approaches,
11 including utilizing zip-code centroids, population-weighted zip-code centroids, and
12 two multiple-imputation methods. Although the main findings here are not
13 particularly sensitive to the selection of particular imputation methods, the analysis
14 demonstrates several alternative approaches with advantages over merely dropping
15 observations with missing data.

16 The results of this study indicate that support for the BeltLine among residents
17 significantly decays along with distance to the project. Yet this study also points to
18 individual desire for accessibility as the main factor behind the distance relationship,
19 which is not expected under the homevoter hypothesis. Also, the results show that
20 residents with kids in affected school zones express less support for the BeltLine.
21 Taken together, this suggests that public support for large redevelopment projects may
22 depend on more than just how the project impacts housing values. Individual tastes
23 and household circumstances play a role as well, which raises more questions about
24 gentrification and spatial sorting as key to driving public support for urban
25 regeneration.

1

2 **2. Background**

3 **2.1 The Atlanta BeltLine**

4 The BeltLine is an urban redevelopment project to transform 22 miles of
5 historic railroad corridors into 22 miles of pedestrian-friendly rail transit and create 33
6 miles of multi-use trails, 1,300 acres of parks, and 5,600 units of affordable housing to
7 connect 45 neighborhoods around the center of Atlanta, Georgia. Stemming from a
8 1999 master's thesis by Georgia Tech student Ryan Gravel, the BeltLine project
9 gained support from Atlanta in 2005, and became an ongoing project after the creation
10 of Atlanta BeltLine, Inc. in 2006. The BeltLine TAD serves as the primary funding
11 source of the BeltLine. 6,500 acres of TAD is projected to generate \$1.7 billion in
12 tax revenue in a twenty-five year window, which is about sixty percent of the original
13 estimated cost. The remainder of the cost is expected to be covered by local
14 contributions and federal funds (Atlanta BeltLine Inc., 2013). Figure 1 illustrates the
15 location of the 22-mile railroad corridors and 6,500 acres of Atlanta BeltLine TAD.

16 **[Insert Figure 1 here]**

17

18 The Atlanta BeltLine project is expected to boost property values in its host
19 neighborhoods and perhaps in an even larger range. Immergluck (2009) conducted a
20 hedonic housing price analysis for single-family house sales in Atlanta from 2000 to
21 2006, and found that after 2005, sales closer to the BeltLine TAD enjoyed a premium
22 in sales price. Immergluck (2009) claims that this proximity premium is a result of
23 both gentrification and local newspaper coverage as speculation bids up prices for an
24 as-yet-unbuilt BeltLine. In 2005 alone, more than 100 stories about the BeltLine
25 appeared in *Atlanta Journal-Constitution*, a major daily paper in Atlanta
26 (Immergluck, 2009). The real estate market in the city may have been broadly

1 impacted in response to the media coverage, though the effect should decay as
2 distance increases.

3 **2.2 Support for Urban Redevelopment Projects**

4 The relationship between support for urban redevelopment projects and
5 property value increments can be described by the homevoter hypothesis. The
6 homevoter hypothesis holds that homeowners politically support actions of local
7 governments that increase property values (Fischel, 2005). The homevoter
8 hypothesis has received some empirical support (e.g., Been, Madar, & McDonnell,
9 2014; McGregor & Spicer, 2014; McLaughlin, 2012). Brunner et al. (2001, 2003)
10 analyze the voting results for a school voucher referendum in California and conclude
11 that homeowners in neighborhoods with superior public schools are less likely to vote
12 for the voucher, because of concerns that property values would decrease. Dehring
13 et al. (2008) also support the homevoter hypothesis in their analysis of the results of a
14 2004 referendum in Arlington, Texas, concerning a publicly subsidized stadium to
15 host the National Football League's Dallas Cowboys. The Atlanta BeltLine case
16 provides an additional chance to indirectly test the homevoter hypothesis. Under the
17 assumption that the level of property value increase caused by Atlanta BeltLine is
18 correlated with distance, as shown by Immergluck (2009), the homevoter hypothesis
19 can be indirectly tested by showing that the distance to BeltLine is correlated with the
20 support.

21 Other factors that correlate with distance might also help explain public
22 support. A household's particular demand for the project's amenities likely
23 correlates with its proximity. Unlike home location, these factors are not capitalized
24 into housing values and thus would not factor into a homevoter's support of the
25 project. Yet if factors other than expected property value impacts influence public

1 support for a project, then households' expected use values may also predict project
2 support.

3 **2.3 Tax Increment Financing**

4 The introduction of TIF is another important factor that can affect support for
5 urban redevelopments. A widely used local government tool for financing economic
6 development in the United States, the main advantage of TIF is providing new funds
7 currently without raising tax rates or providing new revenue-raising authority
8 (Briffault, 2010; Man & Rosentraub, 1998). One concern about TIF related to
9 examining the homevoter hypothesis is the impact of tax-reallocation on education
10 expenditures. Weber (2003) analyzes TIF's impact on the finances of school districts
11 in Cook County, Illinois, and reveals that municipal use of TIF depletes the property
12 tax revenues of schools during the lifespan of the TIF. Part of this is by design,
13 where TIF districts span multiple taxing jurisdictions (Brueckner, 2001). Since the
14 property tax provides a large share of the public school revenue (McGuire et al.,
15 2015), and the quality of public schools is a critical determinant of property values
16 (Brasington, 1999; Haurin & Brasington, 1996), the quality of public schools and
17 property values interact with each other. If the homevoter hypothesis holds,
18 homeowners within the TAD, or those who live outside the TAD but expect the
19 project to lower their school quality and decrease their property values, are likely to
20 favor the BeltLine less. Crucially, this school quality effect is capitalized based on
21 the jurisdiction of the property and not whether the property's resident currently has
22 kids.

23 These concerns over TIF and school quality around BeltLine project are not
24 just theoretical. A year after the TAD was created in 2005, the BeltLine's TIF was
25 legally challenged on the grounds that it unconstitutionally reallocated funds away

1 from the APS. In 2008, the Georgia Supreme Court agreed and ruled the TAD
2 unconstitutional. Later that year, voters approved a statewide referendum to ratify a
3 constitutional amendment that allows using school taxes to fund TADs. By the
4 summer of 2009, this victory for BeltLine proponents resolved considerable legal
5 uncertainty around the project and also raised awareness in the general public of the
6 TIF-related interdependence between the BeltLine and APS. (Uncertainty over the
7 outcome, especially for transit components, remained.) Thus the summer of 2009
8 marked a key juncture in the BeltLine's timeline where the project was approved and
9 implementation had just begun, but its completion was still many years away. The
10 2009 general public survey usefully captures a sort of "baseline" condition
11 uncontaminated by a post-2009 implementation that features notable successes, more
12 legal controversies and disputes over diminished school funding, and growing
13 concerns over gentrification and inequitable roll-out of the BeltLine.

14

15 **Data**

16 Data used in this study are mainly collected from an online survey conducted in
17 the summer of 2009 about the Atlanta BeltLine. At that time, the project
18 construction had begun but very little of it was open. 37 questions were asked of
19 participants' backgrounds, their opinions about the Atlanta region as it was in 2009
20 and as it might become, and their attitudes and expectations about the BeltLine
21 project. To mitigate a social desirability bias (Krumpal, 2013) and BeltLine-related
22 response bias, the invitation letter indicated that it was an opinion survey for Atlanta
23 area residents on the topic of "housing, green space, and transportation." A random
24 sample was drawn from Survey Sampling International's (SSI) online panel, selecting
25 adults in the Atlanta metropolitan area, with 60 percent of respondents from within

1 the city of Atlanta. A response rate of five percent is reported, which is favorable
2 compared to other web-based surveys at the time (Kaplowitz, Hadlock, & Levine,
3 2004; Nulty, 2008). SSI's online panel of 1.5 million panelists has the same age
4 profile as non-panelists but appears somewhat more female and less employed. Our
5 sample closely resembles the 2000 Census statistics for the Atlanta Metropolitan
6 Statistical Area in terms of household size, income, housing tenure, and commute
7 times, although our sample is older and more educated than metro averages. The
8 spatial distribution of all 946 respondents can be shown in Figure 2, and the
9 descriptive statistics of key variables are summarized in Table 1.

10 **[Insert Figure 2 here]**

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15 **[Insert Table 1 here]**

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18 The *Support for BeltLine* variable captures respondents' responses to the
19 question "Do you think that the BeltLine project is a good or a bad idea?" The
20 survey presented five response options: "It is definitely a good idea," "It is more good
21 than bad," "It is more bad than good," "It is definitely a bad idea," and "I need more
22 information to decide." Table 1 reports statistics for *Support for BeltLine* after
23 coding along a -2 to 2 scale (i.e., -2 represents "definitely a bad idea", 2 represents
24 "definitely a good idea"). 18.7% of respondents selected "need more information,"
25 and their *Support for BeltLine* is coded as missing. Respondents are generally
26 supportive for the BeltLine. Only 10% of the sample indicate the project was a bad
27 idea (i.e., *Support for BeltLine* < 0). Respondents also tend to have relatively strong
28 beliefs that the BeltLine will transform Atlanta. This is consistent with proponents'
29 arguments that the BeltLine will transform the whole city. When asked to assume

1 that the BeltLine is completed as planned, respondents show a variety of expectations
2 about the frequency of their future use of BeltLine greenspace and transit. They are
3 also, on average, pessimistic about Atlanta transit and quality of life generally. On
4 average, respondents are 49 years old and have some college education, a household
5 of 2.6 people, and an annual household income of \$68,000.

6 Since this study focuses on the relationship between support of and the
7 distance to the BeltLine, measuring the location of respondents is critical. A final
8 question in the survey asks respondents if they want to receive a report when the
9 survey is done. Half of them provided their specific mailing addresses to receive the
10 report. For the other half that declined, their locations are only known at the zip
11 code level. One simple solution to getting accurate locations is to drop all records
12 lacking precise addresses. But dropping these records will possibly cause two
13 problems. First, dropping half of the records with accurate information because of
14 missing values for a single variable discards a lot of otherwise good information.
15 This is a waste of data from an efficiency perspective. Secondly, and even more
16 importantly, dropping records without precise addresses raises the concern of
17 selection bias. People willing to receive reports may care more about greenspace
18 and transportation and thus are more likely to support projects like the BeltLine. In
19 this sample, those providing street addresses expressed more optimism about the
20 BeltLine as reflected in a greater mean *BeltLine will transform ATL* value (0.95) per a
21 t-test ($t=3.0$).¹ To keep discarding information and to avoid selection bias, this study

¹ While mean *Support for BeltLine* is not significantly different between the sample providing addresses and sample not providing addresses here, other variables do differ between samples. Those reporting addresses tended to be in the City of Atlanta (though not necessarily in the TAD), to be more optimistic about the BeltLine, and to expect to use the BeltLine greenbelt and transit more than other respondents. The means of the other variables in Table 1 are not significantly different between samples.

1 introduces four approaches to impute missing locations as described in detail in the
2 Methodology section.

3 Another critical issue regarding distance is whether the respondent is inside
4 the ring or “donut” of the BeltLine. The role of distance inside the donut is mixed,
5 because moving away from one side of the BeltLine means moving closer to the other
6 side. Thus, the influence of distance for this group of respondents is expected to be
7 smaller than for those outside the ring. The sample size of this group, however,
8 might be too small to affect the overall result. Only seven (out of 459) respondents
9 with actual addresses reside within the donut. After including all missing-address
10 respondents that are in zip codes adjacent to BeltLine TAD, the total possible “donut
11 hole” respondents are only 20 (out of 854). For the simplicity of interpretation, the
12 distance to the BeltLine TAD is logged.

13 The maps of jurisdictions (including the zip code maps, the boundary of the
14 BeltLine TAD and the city of Atlanta) are obtained from the City of Atlanta
15 Department of Planning GIS. The block group-level census data are publicly
16 available from the United States Census Bureau.

17 **Methodology**

18 This study explores the factors related to public support for the BeltLine,
19 especially the distance to the BeltLine TAD. To identify how distance and having
20 kids relate to opinions about the BeltLine, an ordered logistic regression model is
21 estimated:

22

$$23 \text{ Support} = \alpha + (\text{distance})\beta_1 + (\text{jurisdiction})\beta_2 + (\text{kids})\beta_3 + (\text{jurisdiction} \times \text{kids})\beta_4 \\ 24 \quad \quad \quad + (\text{renter})\beta_5 + (\text{renter} \times \text{distance})\beta_6 + X\gamma + u$$

25

1 Where β and γ are vectors of coefficients, X is a vector of other explanatory
2 variables (e.g., household income, age, education, tenure) and u is the error term.
3 The *kids* variable represents the number of kids in the household, which is generated
4 from the survey question about household size. For a household with three or more
5 members, *kids* is defined as the household size minus two. For a household size of
6 two or less, *kids* is defined as zero. As mentioned previously, since lowering future
7 school quality could affect APS-wide property values, the *jurisdiction* dummy is
8 defined as being in City of Atlanta. The interaction term between *jurisdiction* and
9 *kids* is introduced to capture the additional concern of school quality for parents in
10 Atlanta. The dummy and interaction variables for *renter* are used to identify the
11 possibly different distance relationships for homeowners and for renters. If the
12 homevoter hypothesis holds, the *distance* coefficient should differ between
13 homeowners and renters (i.e., $\beta_6 > 0$), because property value changes have different
14 meanings for these two groups.

15 In order to further explore the sources behind the BeltLine's public support,
16 the expected usage of BeltLine amenities are added into the model:

17

$$18 \quad Support = \alpha + (distance)\beta_1' + (demand)\beta_7' + (jurisdiction)\beta_2' + (kids)\beta_3' \\
19 \quad \quad \quad + (jurisdiction \times kids)\beta_4' + X\gamma' + u'$$

20 If the expected *demand* absorbs most of the significance of distance's role, the
21 household's intent to use rather than the property's accessibility is revealed to be the
22 main mechanism behind the distance coefficient. In that case, mechanisms related to
23 the value attached to the property are less important for residents' support. Two
24 variables, the expected frequencies of using BeltLine parks and of using BeltLine

1 transit *once the BeltLine is completed as planned*, are used here to represent the
2 individual demand for key BeltLine amenities.

3 These models identify the factors that help explain BeltLine support or the
4 degree to which the respondent thinks the major urban redevelopment project is a
5 good idea. These factors include various respondent characteristics like income,
6 education, and age, as well as property-related measures that influence support per the
7 homevoter hypothesis. Under that hypothesis, and assuming the BeltLine raises
8 home prices more for nearby homes, the distance coefficient should be negative.
9 Further, that should be true in a model conditional on other respondent characteristics,
10 because those are not capitalized into property values whereas location is. If renters
11 do not benefit from rising housing prices (e.g., Noonan, 2012), the homevoter
12 hypothesis predicts less support from renters (i.e., $\beta_6 > 0$).

13 The negative correlation between *Support* and *Distance* (Spearman rank-order
14 $\rho = -0.16$) still leaves much variation in public support to be explained. The models
15 allow for other factors, beyond property value impacts, to also explain variation in
16 support. In addition to common socio-economic factors like income and education,
17 length of time in residence is included in the first set of models. Newcomers'
18 support might differ, for instance, if they prefer their neighborhoods to remain as they
19 recently selected into or if they moved in anticipation of the BeltLine's
20 transformation. The second set of models explicitly control for individuals' expected
21 use of the future BeltLine, once completed as planned. While this expected use
22 negatively correlates with distance, whether it is location, tastes, or both that explain
23 variation in support remains an empirical question. Because support and expected
24 future use are simultaneously determined by respondents in the survey, causal
25 interpretations are not warranted.

1 As mentioned previously, only half of the respondents provided their
2 addresses. To expand the sample size and to avoid selection bias, this study
3 introduces four approaches to impute missing locations. First, zip code centroids are
4 used to represent the locations of these no-address respondents. This approach has
5 two significant shortcomings. To start, assigning missing-address respondents to zip
6 code centroids brings in measurement error. For zip codes containing large
7 non-residential areas, such as a large park or public facilities, using centroids may be
8 misleading even on average. Moreover, assigning all missing-address respondents to
9 zip code centroids eliminates the potential power of within-zip-code distances to
10 explain different support levels. Introduce this measurement error likely attenuates
11 the distance coefficients toward zero.

12 Second, instead of using simple geographic centroids of zip codes,
13 population-weighted centroids can be generated by overlapping the census block (i.e.,
14 smaller than block groups) population map and zip code map.² This captures the
15 population distribution at the block level within each zip code area. This approach
16 should be more accurate than geographic centroids by taking the within-zip-code
17 population distribution into consideration. Population-weighted centroids can help
18 avoid the first shortcoming mentioned in the previous paragraph. This approach,
19 however, does not help mitigate the problem of eliminating the within-zip-code
20 explanatory power of distance, since all missing-address respondents in a given zip
21 code are still assigned to the same location.

22 The third approach imputes missing distances with all the available variables
23 in the dataset. In practice, the imputation approach first regresses valid distances on
24 all of the other variables (excluding *Support for BeltLine* in this case), and utilizes the

² For a census block overlapping multiple zip codes, the census block is divided into pieces by zip code boundaries. The population of the census block is then distributed by the area of each piece. The population-weighted centroid can thus be generated using the software ArcGIS.

1 regression results to impute missing distances (Little, 1992). This approach
2 generates a specific distance for each missing-address respondents, and thus
3 eliminates the problem of assigning many missing values to the same location. As a
4 result, the distance coefficient with this approach is expected to fit more precisely in
5 the main regression model than those with centroid-based approaches.

6 One concern about the imputation method is that the auxiliary regression
7 coefficients are directly applied to the imputation of missing distances, neglecting the
8 fact that regression estimates (i.e., imputed values) are distributions, not precisely
9 measured values. To fix this problem, this study introduces a multiple imputation
10 approach as the third approach to generating missing distances. The concept of
11 multiple imputation is similar to simple imputation except that it explicitly accounts
12 for the noise in the imputed values. Instead of using fitted values from the auxiliary
13 regression as if they were the measured value, multiple imputation takes a random
14 sample of imputed values based on the estimated coefficient distributions in the
15 imputation regression (Rubin, 1987). Each estimated distance is then used in the
16 main regression. After repeating this imputation-regression process multiple times, a
17 series of regression results is combined into a single set of results. In this study, the
18 imputation-regression process is repeated 100 times.

19 Finally, the fourth approach of filling missing distances applies a truncated
20 regression method to the multiple imputation process. One concern of the
21 imputation process is that the imputed distance might fall outside of the possible
22 range, given the restriction of zip code boundary. For each missing-address
23 respondent, the possible distance to the BeltLine TAD is bound by the shortest and
24 longest distance from the zip code to the BeltLine TAD. To add this restriction to
25 the multiple imputation process, this study introduces the truncated regression

1 method. By providing the lower and upper bounds for each missing distance,
2 truncated regression allows the multiple imputation process to generate imputed
3 distances that are within zip code boundaries.³ Again, the imputation process is
4 repeated 100 times.

5 These four methods for generating missing distances allow the main
6 regression model to be estimated. The estimated coefficients of distance are then
7 compared with each other and with the estimator generated by including observations
8 with actual addresses only (i.e., listwise deletion).

9 The generation of jurisdiction variables is straightforward for respondents with
10 actual locations. Dummy variables are generated with GIS tools, based on whether
11 they are in the jurisdiction or not. It is a more complicated task for missing-address
12 respondents, since their actual locations are not known. In this study, the proportion
13 of zip code area within certain jurisdiction district is used to generate the value when
14 missing. For example, for missing-address respondents in a zip code that does not
15 intersect the BeltLine TAD, their *In TAD* jurisdiction variable is coded as zero. For a
16 zip code that is only partly inside the BeltLine TAD, the *In TAD* jurisdiction variable
17 is coded as the proportion of area overlapping the BeltLine TAD. This same
18 approach to jurisdictional variables is followed regardless of which distance
19 imputation method is used.

20 **Results**

21 Table 2 displays the regression results. The dependent variable measures the
22 individual support for BeltLine on an ordinal scale. The independent variables
23 include: logged distance; jurisdiction dummies (located in BeltLine TAD, City of

³ Due to computation limitations, the upper bound of missing distance is generated by doubling the distance between lower bound and geographic centroid of the zip code:
(Upper bound distance)=(lower bound distance)+2×(Distance between lower bound and the centroid)

1 Atlanta); number of kids; interaction terms between *Kids* and jurisdiction dummies;
2 and demographic characteristics of respondents, such as logged household income,
3 years living in current residence, age, and years of education. Each column
4 represents a specific approach to imputing missing distances. Column 0 lists results
5 for actual-address respondents only. For comparison, column 0' locates all
6 respondents to the corresponding zip code centroids, even if precise addresses are
7 known. Column 1 locates missing addresses at their zip code geographical
8 centroids. The comparison between column 0' and column 1 immediately shows the
9 importance knowing at least some precise addresses for estimating distance
10 coefficients. Column 2 uses population-weighted zip code centroids. Column 3
11 utilizes multiple imputations. Column 4 applies multiple imputations via truncated
12 regressions.

13

14 **[Insert Table 2 here]**

15

16

17 The low p-values shown in the model diagnostics of Table 2 indicate that all
18 the models listed are statistically significant, as compared to the null models with no
19 predictors. The reported pseudo R-squared values are relatively low. However, it is
20 generally perceived that goodness of fit is not as important as statistical significance
21 of explanatory variables (Estrella, 1998; Wooldridge, 2002).

22 The distance coefficient is negative and significant for all four imputation
23 methods, as well as the model without any imputation (column 0). For other
24 variables, imputation generally does not substantively affect the result, no matter
25 which method of imputation method is selected. In other words, imputation enlarges
26 the sample size without disturbing the result. The coefficients for the jurisdictional

1 dummies (TAD, ATL) do change across imputation methods as expected, because
2 different geographic information is used. These dummy variables prove
3 insignificant in the model with the best imputed distance (column 5). Larger sample
4 sizes with more information about distances should not yield similar results as
5 reducing measurement error should lessen the attenuation of the *distance* coefficient
6 toward zero. Comparing the results in column 0', where only zip code centroid
7 distances are used (even if precise distances were known), to column 1 demonstrates
8 this. The larger standard errors for the *Distance* coefficient in Column 0' shows that
9 imputing missing values with limited information in the underlying spatial data can
10 make it harder to detect underlying relationships. Obtaining precise location data for
11 even just a subsample, as was done in this study, enables stronger results and more
12 robust imputation methods.

13 The selection of imputation methods has an interesting effect on the result in
14 terms of both magnitudes of coefficients and their significance. Generally, as more
15 information and more robust imputation is performed, the estimated coefficient for
16 distance grows (more negative) and it approaches the coefficient in column 0 with the
17 smaller sample and no imputation. Imputing distance, however, does not alter its
18 insignificant interaction terms (with *renter* and *newcomer*). The distance
19 coefficients in all four imputation models range from -0.23 to -0.38. For truncated
20 ordered logit model, holding all the other variables constant, increasing the logged
21 distance to the BeltLine TAD by one unit will decrease the ordered log odds of having
22 a higher level of support by 0.38. The jurisdictional variables show mixed results,
23 generally negative for being in the TAD and positive for being in the city, though the
24 estimates are noisy or have somewhat larger standard errors. Geographic location in

1 terms of proximity to the BeltLine offers a stronger predictor of BeltLine support than
2 the geographic jurisdiction indicators.

3 Generally, demographic variables and number of kids alone do not
4 consistently explain different attitudes towards the BeltLine in any models. The
5 variable *Kids* is not significant. In theory, the number of children can affect the
6 attitudes toward the BeltLine in two ways. First, having more children could
7 potentially create additional value from access to parks, trails, and even transit for
8 parents of children who enjoy these amenities. This additional support from parents,
9 however, does not appear in the results. Second, as mentioned previously, parents
10 with kids may worry that the implementation of the BeltLine TIF might hurt the
11 future quality of public schools, thus reducing their support of the project. This lack
12 of support should be sensitive to jurisdictions. Only parents in school zones affected
13 by the BeltLine TAD need to worry about this. Rather than include school
14 catchment zones in the model, where boundaries vary by grade levels, jurisdiction
15 (City of Atlanta) should proxy for school quality effects of the BeltLine as the fiscal
16 impact of the BeltLine TIF will be eventually borne by all public schools in the city.
17 The amenity demand may rise with more children regardless of the jurisdiction, but
18 the concern over school quality impacts should be rising with more children only for
19 those in the APS jurisdiction.

20 The interaction terms between jurisdictions and kids number generally support
21 the argument in the previous paragraph. The interaction terms between *In Atlanta*
22 and *Kids* show a strong and significant negative relationship to support. Holding all
23 the other variables constant, having one additional kid decreases the ordered log odds
24 of being in a higher level of support by 0.84 for respondents in City of Atlanta but not

1 otherwise. Further, the joint significance test for the two *Kids*-related variables
2 shows that they together are related to respondents' support of the BeltLine.

3 The two renter-related variables are not significant (individually or jointly) for
4 all models. Renters' support for the BeltLine is not significantly different from that
5 for homeowners. This result is unexpected in light of the homevoter hypothesis.

6 The homevoter hypothesis holds that property value increases are the main
7 mechanism behind the distance or proximity effect. In this case, renters would not
8 be as supportive of the BeltLine as homeowners at the same close distance, because
9 renters will suffer from the property value increase in terms of higher rents while
10 gaining no benefits from speculating on the as-yet unbuilt project.

11 The fact that renters in the sample do not favor the BeltLine less than
12 homeowners implicitly rejects the homevoter hypothesis. To further confirm this
13 result, a Chow test is introduced. By interacting the variable *renter* with all the other
14 explanatory variables in Table 2, the results for homeowners and renters are estimated
15 separately. The joint F-test for all the renter-interacting variables fails to reject ($F =$
16 0.50) the hypothesis that renters support the BeltLine identically to homeowners. In
17 other words, there is no evidence showing that renters favor the BeltLine differently
18 than homeowners in the sample. Albeit indirectly, the homevoter hypothesis is not
19 supported in this case.

20 To further identify the mechanism behind the role of distance, the second set
21 of models that includes expected use variables are introduced in this study. The
22 results are listed in Table 3. Both restricted models (without expected use) and
23 unrestricted models are listed in Table 3, for comparison purposes. Because the
24 results from different imputation methods are so similar to each other, the comparison

1 in Table 3 focuses on specifications using only multiple imputation with truncated
2 regressions.

3 **[Insert Table 3 here]**

4 The expected use variables are strongly significant in the unrestricted model.

5 Also, the distance coefficient fades after including expected use. This result

6 suggests that the main mechanism behind distance's role is the future accessibility to

7 and expected use of BeltLine amenities. Homevoters should support a BeltLine that

8 raises their property values, an effect related to property distance (see Immergluck,

9 2009), regardless of the tastes or expectations of the property's current resident. Yet

10 Table 3 indicates that it is the current resident's expected use of the BeltLine that

11 drives support, rather than the property's proximity to the BeltLine. There are

12 several explanations for this surprising result. First, the logged distance to the

13 BeltLine TAD might not be a good proxy to the price gradient caused by the project.

14 Given that the BeltLine is a mixed project that includes green space, transit, and

15 affordable housing, the price gradient might not be as straightforward as a function of

16 distance. Noonan (2012) provides some empirical evidence that the price impacts of

17 BeltLine (driven by speculation) are not consistently positive according to a variety of

18 hedonic price models. Second, this study examines survey responses instead of

19 actual votes. The online sample and potentially less deliberation in survey answers

20 may, though not likely, bias the result. Finally, residents might just not be rational

21 or deliberative enough to consider their support for the BeltLine outside of their direct

22 use value. In this regard, the homevoter hypothesis finds little support in the case of

23 the Atlanta BeltLine.

24 **Discussion and Conclusions**

1 This study explores the relationship between distance and other factors in
2 explaining support for the Atlanta BeltLine. Public support significantly declines as
3 distance to the BeltLine increases. The inclusion of expected future use in the
4 model, however, reveals that individual preferences and expected use drive this
5 relationship rather than an independent role of proximity. Under the assumption that
6 property value increases due to the BeltLine rise with proximity to it, these results
7 provide little support to the homevoter hypothesis. The results point to the general
8 public thinking of this major redevelopment project in terms of its costs and benefits
9 to them rather than as a "homevoter" concerned merely about their property values.
10 For example, the results show that parents in the City of Atlanta support the BeltLine
11 less. This supports the conclusion that parents' concerns about TIF affecting future
12 school quality for their kids (rather than for their property values) hampers the support
13 for the project. Such a finding using 2009 survey data comports nicely with renewed
14 legal challenges over financing out of school funds post-2009 and the recent and
15 ongoing conflict over the BeltLine TAD's impact on APS financing.

16 Gentrification around the BeltLine raises interesting and important questions for
17 our understanding of public support for projects like this. The BeltLine
18 redevelopment areas may target neighborhoods populated by those more apt to
19 support it, just as supporters may move nearby in anticipation. This spatial sorting
20 around the project is both an essential aspect of the project's design and success and a
21 source of concern for those seeking equitable redevelopment (Noonan, 2012). The
22 unrepresentativeness of BeltLine neighbors and the prospects of more moving in raise
23 important challenges for measuring the project's costs and benefits, and understanding
24 where its incidence occurs. The neighbors' benefits and costs do not generalize to
25 other populations. Turnover in neighbors, themselves a mix of owners and renters,

1 complicate attempts to identify the project's impacts. While this study examines a
2 snapshot of support in the early stages of implementation, future work would do well
3 to study the dynamics around the BeltLine and similar projects going forward. To
4 date, the BeltLine reports nearly a half billion dollars in investment, seven miles of
5 completed trails, 200 new acres of parks, and 15,400 new housing units (Atlanta
6 BeltLine Inc., 2015). But concerns over progress, affordable housing, and transit
7 (Blau, 2016; Mehrotra, 2014) grow even as the BeltLine grows. How public support
8 shifts, and how the public itself shifts, over time present important aspects of these
9 kinds of projects.

10 The results of different imputation methods shed a light on the technical problem
11 of missing precise spatial location values. Generally, imputation enriches the sample
12 size without altering the results when there are at least some precise location values
13 available. The selection of imputation method does not seem to be a critical issue in
14 this case, since the results remain consistent among methods.

15 The most important policy implication of this study follows from its
16 disconnecting property value increases with public support for the BeltLine. In this
17 case, residents support the Atlanta BeltLine because it provides *them* with local
18 amenities, without separate considerations for this project's impacts on their housing
19 prices. This is an important finding, especially for urban planners who seek for
20 public support. Urban redevelopment projects worldwide usually emphasize on the
21 economic regeneration (Couch & Dennemann, 2000), but local citizens also value
22 facilitation of daily lives, such as public transit (Chan & Lee, 2008). APS parents and
23 would-be users support the project differently even if its impacts capitalize into
24 property values regardless of respondent attributes. This result provides guidance
25 for the promotion of these kinds of urban redevelopment projects. The findings

1 should be interpreted with caution, however, since the support and the usage
2 expectation are decided simultaneously. For example, it is inappropriate to claim
3 that enhancing residents' expectations about their parks usage will stimulate their
4 support for the program based on the result of this study. While both measures move
5 together, the causal relationship remains unsettled in this case.

6 The results shed light on factors associated with supporting this major urban
7 redevelopment project, but other influences remain undetected. Location clearly
8 matters, but residential locations may be driven by BeltLine opinions rather than the
9 other way around. More work is needed to identify other key drivers. Our models
10 of public support show many commonly used measures (e.g., income, age, renter
11 status) as insignificant. The TIF funding approach and the homevoter hypothesis
12 provide some variables to explore, but alternative explanations exist. Factors like
13 race and workplace location, unmeasured in this survey, may also matter. Even
14 though our 2009 'snapshot' analysis of BeltLine attitudes can say little about
15 gentrification, survey responses foreshadowed the renewed conflict over school
16 funding (Blau, 2016) and the BeltLine's popularity today among trail users (Mehrotra,
17 2014). As other cities like New York and Seoul implement major, transformative
18 projects, researchers would do well to learn from those cities' experiences to better
19 understand the nature and drivers of public support for the redevelopment.

20 There are several concerns and limitations to this study from the data analysis
21 perspective. The first issue is the possible measurement error issue in the dependent
22 variable: the support for the BeltLine. Respondents of the survey hail from all over
23 the metro Atlanta area. It is arguable that some of them are too far away to credibly
24 express support for the BeltLine— though the effective range of BeltLine is also
25 arguable. One argument is that these distant responses are just random noise, which

1 should not affect the estimates. Another argument is that these distant respondents
2 only respond because they care about the BeltLine, even though they are not really
3 affected or should otherwise be excluded from the study. In this case, the support
4 from these distant observations in the sample are expected to be “too high” compared
5 to the population. This self-selection problem would result in underestimating the
6 distance effect, because the support in distant areas is not as low as it should be. In
7 sum, the measurement error caused by including excessively distant respondents
8 either does not affect the estimators or gives us lower-bound estimates, depending on
9 how the measurement error is interpreted.

10 Another measurement error issue comes from missing addresses. Half of the
11 respondents opted not to provide their addresses when asked if they want to receive
12 the final report of the survey. This missing-address issue is not random, because
13 respondents who decline to receive the report are likely to care less about the topic
14 and be less supportive of the BeltLine. Therefore, dropping less-supportive
15 observations may bias the estimated distance coefficient toward zero, since the effect
16 is diluted. This bias is not evident in the results. The four approaches to impute
17 these missing addresses are not perfect, but offer advantages in terms of avoiding bias
18 while conservatively allowing for additional error in the imputation.

19 Measurement error in the *Kids* variable also needs mentioning. First, using a
20 categorical family size question to infer the number of kids creates error. Again, this
21 error in independent variable is expected to increase standard error and biases the
22 estimator toward zero. Second, not having information on the age of the kids is
23 another limitation of the study. For respondents with older kids, having kids should
24 not affect the distance coefficient, since they care less about the potential influence of
25 the BeltLine on school expenditures. This could attenuate the estimated *Kids*

1 relationships, again providing a conservative estimate here. In addition, failing to
2 account for childless households expecting to have kids in the near future might also
3 weaken the *Kids* coefficients relative to the households without kids. Better
4 measures for household composition in the original survey would have strengthened
5 the results.

6 Generally, the possible sources of measurement errors are likely to either
7 amplify the standard errors or perhaps even bias coefficients toward zero, making the
8 estimators conservative in their approach. Understanding the possible error sources
9 and the consequences helps us appreciate what the true values would likely be, even
10 though the estimators are not perfectly accurate. These results indicate that the
11 considerable public support for the BeltLine can be partly explained by individual
12 respondents' intended future uses of the BeltLine amenity, more so than the distance
13 of their home to the BeltLine. This is true for both owners and renters and
14 regardless of how the imprecision in measuring distance is addressed. Further,
15 support is lessened for households with many kids inside the BeltLine TIF zone's
16 school district, consistent with concerns over future school funding. Altogether,
17 these results show support stemming from respondents' particular circumstances and
18 inclinations more than from the expected impact on their housing values as the
19 homevoter hypothesis would hold.

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