

1 **TRANSIT-SUPPORTIVE MUNICIPAL POLICIES AND STATION-LEVEL**
2 **RIDERSHIP**

3 **A Direct Ridership Modelling Analysis**

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38 **ABSTRACT**

39 Transitways represent large public investments whose positive impacts must be maximized
40 whenever possible to justify the expenditures they entail. Prominent among those looked-for
41 positive impacts is the encouragement of automobile-to-transit mode shifts by attracting
42 increased transitway ridership. Transitways have a ridership impact on their own, but their
43 stations exist in an environment of municipal policies which may not aim to maximize transitway
44 use, and often aim to maximize automotive mobility. This paper explore the impact of municipal
45 policies on transitway ridership through a national study of 16 metropolitan areas to answer the
46 research question: What role and transit-supportive public policies implemented around
47 transitway stations in boosting transitway ridership? The authors employ a Direct Ridership
48 Model (DRM) to predict boarding at the station level. Station-area focused policies promoting
49 affordable housing and explicit policies calling for sidewalks on all streets in station areas or
50 entire cities have a significant and positive impact on ridership if there are sufficient potential
51 destinations in the immediate station area. The results stress the importance of station area
52 affordable housing as a transit system efficiency measure, as well as for the social equity reasons
53 it is usually encouraged. The authors recommend the strengthening of pro-affordable housing
54 policies and pro-sidewalk policies in station areas, support and encouragement for the
55 neighborhood-scale commercial development that is required for their efficacy and the continued
56 implementation of pro-affordable housing policies and pro-sidewalk policies.

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63 INTRODUCTION

64 Despite their recent growth, transitways operate in an overall public policy environment that still
65 strongly favors the automobile (1, 2). This policy environment makes encouraging mode shifts
66 from automobile to transit unnecessarily difficult. Policies ranging from zoning ordinances to
67 off-street parking requirements often date from an era when the unimpeded movement of
68 automobiles was the primary goal of transportation planning. That goal has changed—supplanted
69 by a variety of goals ranging from regional accessibility to vehicle-miles traveled (VMT)
70 reduction to increased transit use. The major investments represented by transitways may call for
71 a reevaluation of existing public policies in station areas in the interest of promoting mode shifts
72 from automobiles to transitways.

73 This report focuses specifically on policies local governments may implement in station
74 areas intended as “force multipliers”, so to speak—enhancements to the basic ridership attraction
75 effects of transitways themselves. To compare systems in broadly comparable metropolitan
76 areas, the research focuses on a set of peer regions defined by the Twin Cities Metropolitan
77 Council (Hiniker, 2013, unpublished material). The authors obtained boarding data for each of
78 these systems shown at the station level, and estimate a direct ridership model to explain
79 ridership as a function of surrounding transit-supportive policies, controlling for transit mode and
80 service characteristics, as well as station area built environment and social characteristics.

81 LITERATURE REVIEW

82 Traditionally, transitway ridership forecasting efforts have depended on the classical, four-step
83 travel demand modeling process, in widespread use since the 1950’s. Four-step models
84 effectively treat trip generation, trip distribution, mode split and route assignment as separate
85 processes. This approach considers the impacts of earlier steps on later steps, but fails to account
86 for feedback loops from later steps to earlier steps—for example, how congestion (considered in
87 the route assignment step) can impact future development decisions (considered largely in the
88 trip generation step) (3). The Transportation Analysis Zones (TAZ’s) used as the units of
89 analysis in four-step models are also—in many cases—too geographically large to effectively
90 analyze walking trips. In addition, four-step models generally do not consider local land use and
91 built form variables (which can be crucial determinants of transit use), and are incapable of
92 accounting for the impacts of residential self-selection—such as occurs when transit users choose
93 to live in TOD’s (3, 4).

94 Direct Ridership Models (DRM’s), also known as off-line models, have recently come to
95 prominence as an alternative. DRM’s predict ridership in a single step using multiple regression
96 analysis, based on regional, station-area and proposed transit service characteristics (3, 5). In a
97 DRM, any factor with an empirical relationship to transit use can be used to predict ridership (3,
98 6). DRM’s require actual transit ridership data, but do not require that all aspects of the utility of
99 choosing a given transit option be explicitly considered—or even known. For example, a DRM
100 can account for the frequently observed, yet incompletely understood tendency of rail transit
101 lines to attract somewhat higher ridership than otherwise similar, bus-based services. The “rail
102 bias” can be implicitly considered by the inclusion of ridership data and modal variables from
103 rail and bus transit modes. Lacking the trip origin-destination specificity of other modeling
104 techniques, DRM’s are generally considered sketch-planning tools—used for a first look at
105 proposed lines or to weigh the relative impacts of different options (3). Within the bounds of
106 these limitations, a growing body of research shows strong potential for DRM’s to predict the
107 ridership of proposed transitways based on realized physical, social and economic

108 characteristics. Lane, DiCarlantonio and Usvyat, in a broad analysis of existing commuter rail
109 and light rail lines in six regions of the United States, found their DRM approach predicted
110 actual ridership with a high degree of accuracy (7). Even so, little research to date has employed
111 direct ridership modeling to analyze the ridership impacts of public policies—such as TOD-
112 friendly zoning, low minimum parking standards (or maximum parking standards), parking
113 pricing, employer-provided transit benefits, and bicycle and pedestrian infrastructure provision—
114 thought to promote transit use.

115 Direct ridership modeling allows ridership predictions for proposed transitways on the
116 actual ridership of well established, existing lines (3). In 2003, Walters and Cervero were able to
117 predict station-level ridership for a proposed extension of the Bay Area Rapid Transit (BART)
118 system in the San Francisco Bay area. By including both existing BART rapid transit stations
119 and Caltrain commuter rail stations, they were also able to predict the ridership impacts of
120 different rail technologies (8). Kuby, Barranda and Upchurch were able to estimate a single
121 model to explain average weekday boardings for 268 light rail stations in nine markedly different
122 cities across the United States. Notably, their results pointed to land use, regional accessibility
123 and bus connections as important predictors of ridership, along with socioeconomic factors,
124 park-and-ride access and regional centrality. Their model achieved an impressive R^2 value of
125 0.727, demonstrating the power of the DRM approach (6).

126 **Transit-Supportive Public Policies**

127 The transportation and urban planning literature broadly recognize the strong influence of factors
128 other than transit service and frequency characteristics on transit use patterns. In a study of 265
129 urban areas across the United States, Taylor, Miller, Iseki and Fink find that a majority of
130 observed variation in per capita transit use rates can be explained by a mix of regional
131 geography, metropolitan economy, population and highway system characteristics. In other
132 words: most of the variation in transit use can be accounted for without even considering transit
133 systems themselves (9). Some of the variables considered in this study—such as total
134 metropolitan population or percentage of Democratic voters—are beyond the control of
135 municipal and regional policy. However, the study also suggests several non-transit system
136 variables that are directly influenced by local and regional public policy, such as population and
137 employment density in transit service areas, and parking supply (9).

138 *Transit-Oriented Development*

139 Attempts to increase transit use through means other than transit system and/or service changes
140 frequently revolve around Transit Oriented Development (TOD). TOD seeks to retrofit transit-
141 friendly built forms onto previously automobile-dominated areas, generally through some
142 combination of high-density housing development, mixed land uses and connected, pedestrian-
143 friendly street environments (10-12). The potential for TOD to increase transit mode shares—
144 along with important limitations to that potential—has been established for decades. Based on
145 early 1990's data, Cervero found that San Francisco Bay Area residents of dense, compact
146 TOD's were more likely than others to commute by rail, but that "The strongest predictors of
147 whether station-area residents commuted by rail was whether their destination was near a rail
148 station and whether they could park for free at their destination" (13). Support for the
149 effectiveness of alternative development patterns' ability to alter travel behavior is not
150 unanimous—Giuliano finds that land use policies are incapable of major travel-behavior impacts
151 (14), while Crane finds the evidence inconclusive (15)—but the bulk of the literature appears to
152 find some positive evidence.

153 Most research on TOD has traditionally been conducted at the local level, focusing on
154 individual developments or groups of developments. Some recent research also considers TOD
155 at the regional level. The importance of transit-accessible destinations is echoed by Tilahun and
156 Fan in a 2014 study of future growth scenarios in the Twin Cities region. Their research found
157 that concentrating future housing development in transit-served areas would have a significant,
158 positive impact on transit-based employment accessibility (thus making the transit system more
159 useful), but that concentrating future employment growth near transit would have a larger
160 positive impact and that the greatest benefit would come from concentrating both (16).

161 Public sector policy decisions have an important role to play in promoting TOD. A
162 variety of researchers find latent, pent-up demand for TOD, along with evidence that
163 automobile-oriented zoning and other development regulations hinder developers from pursuing
164 transit oriented projects (17-20). In addition, Levine and Frank find a relative undersupply of
165 compact, transit-friendly development in the highly-sprawled metropolitan Atlanta region when
166 compared with metropolitan Boston based on residents' preferences (21). In recent research on
167 Twin Cities region, Fan and Guthrie find regulatory structures designed to limit density, separate
168 uses and ensure universal automobile access to be an important obstacle to TOD (17).

169 *Parking Policy*

170 Parking supply and/or pricing can have major effects on travel behavior, and are also in large
171 part determined by public policies. Near-universal minimum off-street parking requirements
172 designed primarily to ensure uninterrupted motor vehicle access (rather than to ensure optimal
173 transportation system performance or highest and best use of urban land, etc.) mean that
174 motorists enjoy free parking for 99% of trips—a subsidy estimated to be greater than free fuel
175 (22). Willson finds that excessive workplace parking requirements impact urban form, transit
176 demand and ultimately transit service. Willson also finds potential for transit providers to benefit
177 from helping shape reformed parking standards (23). Shiftan and Golani also find, based on
178 survey data, that downtown parking reductions and price increases, along with other constraints
179 on automobile use would be more likely to change downtown-bound travelers' mode choices
180 than their destinations (24, 25).

181 The literature suggests that the interrelationship of parking policy and automotive travel
182 demand is often not well understood and/or seen as important by the local planners who craft
183 parking policies and implement them on a project level (22). In a survey of planners from 138
184 California local governments, Willson finds “that the most common response to a question about
185 workplace parking issues was that there were *no* important issues. The next most frequent
186 response was parking *undersupply*.” (Emphasis in original.) Planners gave these responses
187 despite significant empirical evidence that an oversupply of parking is the most common
188 situation (23). The same survey also finds that planners most commonly cite “ensuring an
189 adequate supply of parking” as the primary consideration in parking policy, demonstrating a
190 general lack of critical consideration of the transportation system impacts of parking policy.
191 Willson further finds that the risks of under-building parking (such as neighborhood spillover,
192 congestion, perceived excessive density, etc.) tend to be more clearly perceived by planners (and
193 neighborhood groups) than the risks of overbuilding parking (such as foregone tax revenue,
194 excessively low density, poor urban design and pedestrian environment, increased VMT and
195 difficulty in providing transit) (23). To deal with oft-raised concerns over neighborhood spillover
196 parking, Shoup suggests the creation of “parking benefit” districts, in which revenues from
197 market-priced on-street parking would be directed to benefit the local community. Shoup
198 contends that revenues from fair-market pricing of curbside parking could exceed those from

199 residential property taxes (26). These findings suggest significant room for public-policy
200 improvement in crafting parking requirements with the goal of transit-friendly communities in
201 mind.

202 *Employer Commute Benefits*

203 Employers' benefits packages can also influence commuters' mode choices. Particularly for
204 downtown workplaces—often those for which transit is most likely to offer a viable commute
205 option—employers frequently provide significant parking subsidies which insulate individual
206 employees from the full cost of driving to work. As far back as 1992, Shoup and Willson
207 proposed requiring employers that provide parking benefits to also offer employees a transit pass
208 or cash commute allowance of equal value. They predicted that implementing such a policy in
209 downtown Los Angeles would reduce automotive commuting by 17% while leading to a 67%
210 increase in transit use (27). Acting along similar lines, by offering both employees and students
211 subsidies for carpooling and transit passes as an alternative to parking passes, the University of
212 Washington achieved a 16% reduction in motor vehicle trips to campus, along with a 35%
213 increase in transit trips and a 21% increase in carpools (28). Dill and Wardell find that employers
214 in Portland, Oregon who participate in TriMet's discounted employee pass program have transit
215 mode shares seven percentage points higher regardless of location, twelve percentage points
216 higher downtown and five percentage points higher outside downtown, even when controlling
217 for proximity to LRT and bus stops, employer characteristics and built form characteristics (29).
218 Su and Zhou reach generally similar conclusions for transit benefits in Seattle, Washington, as
219 well as significant impacts for higher SOV (and lower HOV) parking prices (2).

220 *Walkability/Bikeability*

221 Street environments surrounding transit stations can have profound impacts on transit use
222 behavior as well. In a study of Saint Louis Metrolink riders, Kim, Ulfarsson and Hennessy find
223 that perceived security as well as availability and convenience of bus connections had significant
224 impacts on light rail users' access modes (30). Schwanen and Mokhtarian found that
225 neighborhood environmental factors (such as land use mix, density and pedestrian environment)
226 had a greater impact on commute mode choice than residents' neighborhood type preferences,
227 especially in transit-unfriendly environments. According to their research, "Although
228 mismatched suburban residents may be more inclined to use transit than their matched neighbors,
229 many may feel they have no choice [...] but to commute by personal vehicle" (31). Cervero finds
230 that local neighborhood-level land use and urban design factors have significant impacts on
231 mode choice—even controlling for modal travel-time differences. In a study of household travel
232 survey data from Montgomery County, Maryland, density, mixed uses and the prevalence of
233 sidewalks all predicted significantly higher rates of transit and nonmotorized travel (32).

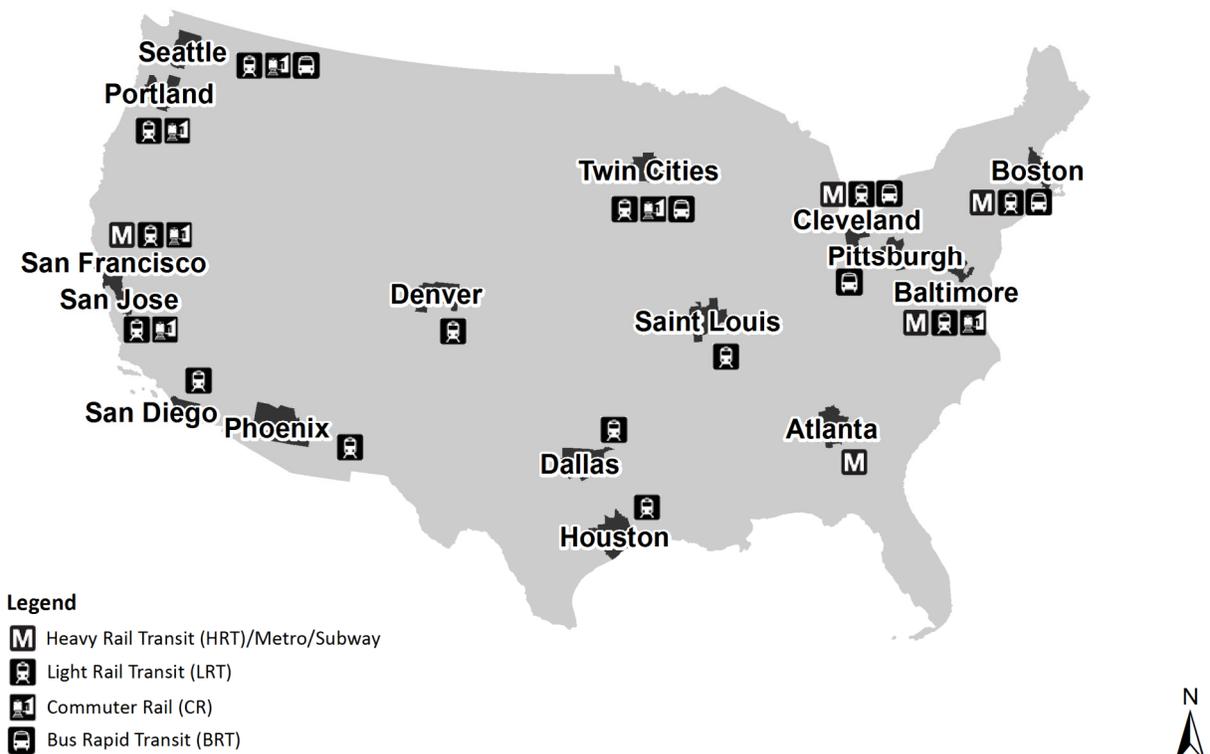
234 Cervero also contends that improving pedestrian and bicycle conditions around
235 transitway stations could have important long-term mode shifting impacts even if the immediate
236 effect is primarily a shift in transit users' access modes from park-and-ride to walk/bike-and-ride.
237 Reduced demand for park-and-ride access would allow transit providers to redirect resources
238 from parking provision to actual transit operations (leading to more attractive services), and
239 reduced use of station area land for parking would ease the promotion of transit-oriented
240 development (leading to increased trip origin and destination densities in station areas) (33).

241 **Summary**

242 Encouraging mode shifts from automobiles to transitways through policies that look “beyond the
 243 rail” is a complex problem which traditional travel demand modeling techniques are ill-equipped
 244 to solve. Multiple areas that are already heavily influenced by local public decision-making—
 245 including transit-oriented development, zoning, parking policy, employer commute benefits and
 246 pedestrian- and bicycle-oriented design—appear to present significant opportunities to encourage
 247 just such mode shifts if the policies guiding them can be coherently retooled with mode shifting
 248 as a goal. Indeed, a review of the literature reveals enough public policy hindrances to auto-
 249 transitway mode shifts that achieving the full mode shifting potential of transitways may simply
 250 be impossible without beyond-the-rail policy tools. Despite the complexity involved, direct
 251 ridership modeling techniques offer a robust method for analyzing the ridership impacts of local
 252 public policy decisions based on actual results from existing, well established transitways..

253 **METHODS**

254 The research approach hinges on the use of station-level boarding data to allow for a model
 255 sensitive to localized, station area conditions. Unlike provider- and mode-level ridership data, no
 256 national source of station level boarding data exists. As such, the first phase of data collection
 257 involved contacting transit providers in study regions to request data. The authors enjoyed an
 258 extraordinary degree of cooperation in this phase: every provider contacted eventually provided
 259 ridership data. Figure 1 shows the regions and modes included in the research.



260

261 **FIGURE 1 Study regions and modes**

262 Ridership Data Issues

263 Despite the willing assistance received from transit providers, several problems with ridership
264 data arose. First of all, there is no universal standard for measuring the ridership of fixed-
265 guideway transit services. Light rail, bus rapid transit and commuter rail systems tend to track
266 average weekday boardings—the number of passengers who board a train or bus at a given stop
267 on an average weekday. Ridership may be counted directly using Automatic Passenger Counters
268 (APC's) or estimated using manually conducted sample counts. If performed properly, either
269 collection method provides a relatively close measure of unlinked trips. Heavy rail rapid transit
270 systems, however, tend to track station entries. These fully grade-separated systems generally
271 use barrier-based fare collection systems (such as turnstiles), with neither additional fares nor
272 transfers required between lines. As a result, this method of ridership data collection measures
273 linked trips. For example: a Twin Cities transit user wishing to travel via light rail from the
274 University of Minnesota to MSP International Airport would board a westbound Green Line
275 train on campus (one boarding), alight at Downtown East and board a southbound Blue Line
276 train (two boardings). A MARTA user in Atlanta wishing to travel from Georgia State
277 University to Hartsfield-Jackson International Airport, however, would enter the system at
278 Georgia State Station (one station entry), board a westbound Blue or Green Line train, alight at
279 Five Points and transfer to a southbound Gold or Red Line train without ever leaving the station.
280 As a result, only the initial station entry is recorded in ridership data, producing an undercount of
281 ridership at transfer stations relative to the methods used by LRT, BRT and commuter rail.

282 Also, in addition to a number of modern “New Start” light rail lines, this study includes
283 four regions with legacy light rail systems converted and upgraded from survivors of their cities’
284 original streetcar systems—Boston (Green Line), Cleveland (Shaker Rapid), Pittsburgh (the T)
285 and San Francisco (Muni Metro). These systems are hybrids—with full grade-separation and
286 subway- or light rail-style stations in the central business district, and more or less traditional
287 streetcar operations further out. Due to somewhat older vehicles and large numbers of
288 rudimentary median or even curbside stops, only the Massachusetts Bay Transportation
289 Authority and the Greater Cleveland Regional Transit Authority were able to provide reliable
290 boarding data for their entire light rail systems. San Francisco Muni was only able to offer
291 station entries for their Market Street subway stations, while the Port Authority of Allegheny
292 County in Pittsburgh does not track light rail ridership at the station level.

293 Policy data

294 As with station-level ridership, no national data source for transit-supportive public policies
295 exists. In fact, even individual municipalities seldom have a single inventory of such policies. As
296 a result, transit supportive public policies required original data collection. Data collection
297 initially took the form of an online survey distributed to planning staff in cities containing study
298 stations. Despite multiple contacts and reminders, the authors failed to obtain a sufficient
299 response rate (Less than one fifth of the cities in our other data) from the survey.

300 To allow the research to proceed, the authors manually coded remaining cities by using the
301 original survey instrument as an audit conducted by a member of the research team. The survey
302 instrument collected data on:

- 303 • Land use policy
- 304 • Pedestrian/bicycle infrastructure policy
- 305 • Street design standards
- 306 • TOD funding/promotion policy

- 307 • Affordable housing policy
- 308 • Transit system fare policy
- 309 • Policy implementation mechanisms

310 To obtain clear, unambiguous answers, questions were phrased as a binary policy present
 311 versus policy absent choice, for example: “Does the city allow higher densities in stations
 312 areas?” or “Does the city have policies promoting the construction of new affordable housing in
 313 station areas?”. Policies focus on basic standards for public facilities (such as calling for
 314 sidewalks along all streets) are considered present if they apply either specifically to station areas
 315 or generally to the city as a whole. Policies dealing with the allocation of public resources or
 316 types of development (such as affordable housing promotion or zoning) are only considered
 317 present if they focus specifically on station areas. For example, a city wide policy requiring
 318 sidewalks on all streets would be coded as present, as would a policy promoting affordable
 319 housing in station areas, while a general, city wide policy of promoting affordable housing would
 320 be coded as absent.

321 **Google Places**

322 The authors also downloaded “places” data in bulk from Google Maps. Google places are the
 323 source data for the business, institution and other destination placemarks that appear in Google
 324 Maps when zoomed in to a relatively small geographic extent. Specifically, the authors
 325 downloaded counts of places, both in total and by type within a 100M radius of each transitway
 326 station included in the data. The authors employ a relatively small radius because Google sets a
 327 bulk download limit of 60 places per central point/station; a larger radius would artificially
 328 reduce variation in the count of places by pushing more stations past 60 places. Google places
 329 provide a measure of neighborhood activity with much finer geographic scale than is available
 330 from other data sources.

331 **Approach**

332 The authors propose that transitway ridership, at the station level, is largely locally determined
 333 by realized station area conditions, such as the built environment and demographics. However,
 334 there also exist a wide variety of public policies which seek to influence those realized
 335 conditions, such as zoning codes, street design guidelines, housing policies, etc. These policies
 336 may impact transitway ridership indirectly by influencing the realized station area conditions
 337 which in turn influence ridership. There may also be an interaction effect between policies and
 338 station area conditions by which the presence of a consistent policy reinforces the impacts of
 339 transit-friendly station area conditions or mitigates the impact of transit unfriendly conditions.

340 The research employ a Direct Ridership Model (DRM) to predict transitway boardings at
 341 the station level as a function of transit-supportive public policies while controlling for transit
 342 service and regional characteristics and station area built environment and social characteristics.
 343 As described by Cervero (2006), DRM’s employ multiple regression to predict ridership in a
 344 single step based on any set of predictors the modeler deems appropriate or is interested to
 345 explore the specific effects of.

346 **RESULTS**

347 Table 1 shows the results of the ridership model. The authors initially estimated a combined
 348 model including all 850 observations. Suspecting an interaction effect between Parking policy
 349 and Google places due to counter-intuitive results, the authors divided the data just below the

350 **TABLE 1 Ridership models**

	<i>Response Variable:</i> Boardings <i>Explanatory</i>	Observations Pseudo R²	<i>Combined</i>		<i><50 Est.</i>		<i>≥50 Est.</i>	
			<i>β</i>	<i>IRR</i>	<i>β</i>	<i>IRR</i>	<i>β</i>	<i>IRR</i>
		850			406			444
		0.74			0.67			0.78
<i>Policies</i>	Pro-sidewalk policy (binary)		0.2289	1.2572	0.2462	1.2791	0.2461 *	1.2790
	Pro-density policy (binary)		-0.0575	0.9441	0.1204	1.1279	-0.2117	0.8092
	Pro-affordable housing policy (binary)		0.2861 **	1.3312	0.0915	1.0958	0.3794 ***	1.4615
<i>Built Environment</i>	Number of establishments (count)		0.0109 ***	1.0109	0.0118 **	1.0118	0.0549 **	1.0564
	Residential density (housing units/acre)		-0.0026	0.9975	-0.0017	0.9983	-0.0046	0.9954
<i>Station Characteristics</i>	HRT (binary)		1.0425 ***	2.8362	1.0837 ***	2.9555	1.0066 ***	2.7363
	BRT (binary)		-0.8989 ***	0.4070	-1.1419 ***	0.3192	-0.8790 ***	0.4152
	CR (binary)		-0.5180 *	0.5957	-0.6940 **	0.4996	-0.2305	0.7941
	Legacy system (binary)		-0.0546	0.9469	-0.3465	0.7072	0.1013	1.1066
	Number of lines (count)		0.3519 ***	1.4218	0.4300 ***	1.5372	0.2879 ***	1.3337
	One-direction station (binary)		-0.7969 ***	0.4507	-0.8327 ***	0.4349	-0.7360 ***	0.4790
	Terminal station (binary)		0.5000 ***	1.6487	0.7391 ***	2.0941	0.2763 **	1.3183
	Park and ride (binary)		0.1269	1.1353	0.1806 *	1.1979	-0.0888	0.9150
	Distance from CBD (kM)		-0.0188 ***	0.9814	-0.0121 ***	0.9880	-0.0234 ***	0.9769
	Distance from next station (kM)		0.0816	1.0850	0.0497	1.0509	0.1490 *	1.1606
<i>Social Characteristics</i>	Median income (\$1K)		-0.0004	0.9996	0.0000	1.0000	0.0003	1.0003
	% 1-person households		-0.1537	0.8575	0.6341 *	1.8853	-0.3974 **	0.6721
	% population < 18 yr.		-1.4274 ***	0.2399	-0.7290	0.4824	-2.1035 ***	0.1220
	% population ≥ 65 yr.		-0.5324 *	0.5872	-1.5189 **	0.2190	-0.2882	0.7496
	% minority population		-0.1718	0.8421	-0.1603	0.8519	-0.2133	0.8079
	% foreign born residents		0.9561 ***	2.6016	0.9316 *	2.5386	1.0602 ***	2.8871
	% households where workers > cars		1.0443 ***	2.8415	1.2660	3.5467	0.8810 ***	2.4133
	Constant		6.1176		5.5933		3.8157	

*p < 0.1; **p < 0.05; ***p < 0.01

351 median of 54 places into two models: one for stations with less than 50 Google places in its
 352 immediate environs and one for stations with greater than or equal to 50 such places.

353 The final models employ Poisson regression, with boardings as the response variable and
 354 the following explanatory variables:

- 355 • Pro-sidewalk policy—Dummy variable indicating the presence of a policy
 356 explicitly calling for sidewalks on all streets either in station areas or generally.
- 357 • Pro-density policy—Dummy variable indicating the presence of a policy allowing
 358 for higher residential and/or commercial densities in station areas than elsewhere in the city. To
 359 yield a value of 1, the policy in question must have a specific station area focus.
- 360 • Pro-affordable housing policy—Dummy variable indicating the presence of a
 361 policy calling for the preservation of existing affordable housing and/or the construction of new
 362 affordable housing in station areas. To yield a value of 1, the policy in question must have a
 363 specific station area focus.
- 364 • Number of establishments—Count of Google places within a 100M radius of the
 365 station. Included as a measure of activity in the station area.
- 366 • Residential density—The residential, in dwelling units per acre, of the census
 367 block group containing the station.
- 368 • HRT, BRT, CR—Dummy variables identifying the transitway mode serving the
 369 station. LRT is omitted as the reference.
- 370 • Legacy system—Dummy variable identifying stations on a legacy rail system
 371 operated by a transit provider that never abandoned rail transit. For example, San Francisco's
 372 Muni light rail and Caltrain commuter rail services are legacies; the Bay Area Rapid Transit
 373 system (BART) is not. Included due to a frequent lack of self-conscious transit-supportive
 374 policies in cities with legacy rail systems, due to transit-oriented environments being
 375 unremarkable in such places.
- 376 • Number of lines—The number of transitway lines serving the station.
- 377 • One direction station—Dummy variable identifying stations only offering service
 378 in one direction, generally where a transitway employs a one-way pair alignment.
- 379 • Terminal station—Dummy variable identifying the terminal of a transitway.
- 380 • Park-and-ride—Dummy variable identifying stations with commuter parking.
- 381 • Distance from CBD—The airline distance, in kilometers, from the station to the
 382 transitway station closest to the heart of the central business district. Included as a measure of
 383 centrality while providing more information on metropolitan radius than a simple city/suburb
 384 dummy variable.
- 385 • Distance from next station—The airline distance, in kilometers, from the nearest
 386 other transitway station. Included to account for station-level ridership differences which may
 387 appear due to the division of transit demand among more or fewer stations.
- 388 • Median household income, % Single-person households, % of residents under 18,
 389 % of residents age 65 and over, % minority population, % foreign born, % of household where
 390 workers > cars—Measures of social conditions in the census block group containing the
 391 transitway station.

392 In the model for stations with less than 50 establishments in their immediate
 393 surroundings, none of the policy variables are significant, indicating there is a minimum level of
 394 station area activity required for the transit-supportive policies considered to have a discernable
 395 effect on transitway boardings. The count of nearby establishments is significant and positive,

396 indicating station area activity has a positive relationship with ridership even without the critical
397 mass needed for transit supportive policies to have significant effects.

398 Nearly all the transit station characteristics variables are at least marginally significant,
399 including park-and-ride, which was insignificant in the combined model. While HRT has a
400 strongly positive coefficient, BRT and commuter rail are both strongly negative. (All modes are
401 compared to LRT.) BRT has a particularly strong negative coefficient, attracting less than one
402 third as many boardings per station as LRT, all else equal. In addition, the model predicts that
403 ridership drops roughly 1.2% for every kilometer of distance from the central business district.

404 Of the social variables included, only percent one-person households, percent of the
405 population age 65 and older and percent foreign born residents are significant. In a notable
406 difference from earlier models, percent of households with more workers than cars is
407 insignificant.

408 In the model for stations immediately surrounded by at least 50 establishments, a pro-
409 sidewalk policy is marginally significant and a specifically station-area focused pro-affordable
410 housing policy is highly significant. Both have relatively strong positive coefficients, as well. A
411 pro-sidewalk policy yields a 28% increase in boardings, while a station-area focused pro-
412 affordable housing policy yields a 46% increase. The number of establishments is still
413 significant, and actually has a more strongly positive coefficient, lending support to the idea of
414 an interaction effect between station area activity and transit-supportive policies.

415 Only HRT (with a strong, positive coefficient) and BRT (with a strongly negative
416 coefficient) are significant among transit modes. The insignificance of commuter rail may stem
417 from high-activity commuter station areas tending overwhelmingly to be located in central
418 business districts, with consequently heavy use. Park-and-ride is also no longer significant as a
419 predictor of ridership, indicating that the absence of a park-and-ride facility is not a significant
420 detriment to transitway use in station areas with sufficient surrounding activity.

421 Among the social variables, percent of one person households remains significant and
422 positive. Percent of the population under age 18 is now significant, with a strongly negative
423 coefficient potentially reflecting lower ridership in single-family residential areas. Percent
424 foreign born residents remains significant and strongly positive. In this model, percent of
425 households with more workers than cars is highly significant, with a strongly positive
426 coefficient. The model shows that every percentage point increase in households with low
427 automobility yields a 2.4% increase in transitway ridership.

428 Table 2 shows standardized coefficients for the final policy model for station areas with
429 at least 50 establishments. The strength of the two significant modal dummy variables (positive
430 for heavy rail, negative for BRT) is striking. These variables compare their respective modes to
431 light rail. It is likely that some degree of modal bias is reflected in the results. However, one must
432 bear in mind that the relationship between mode and ridership may simply reflect the selection of
433 transit technologies based on predicted demand. A common pattern among regions studied is to
434 use rail for their highest-demand corridors, complemented and augmented by BRT on
435 corridors with more moderate demand—a practice which reflects Twin Cities transitway plans as
436 well.

437 Another striking result is the relative strength of a station area focused pro-affordable
438 housing policy. It is the fourth strongest predictor of ridership in the model, stronger even than
439 the number of households with more workers than cars. Pro-affordable housing policy is also
440 significantly stronger than pro-sidewalk policy.

441

442 **TABLE 2 Standardized coefficients, ≥ 50 establishments model**

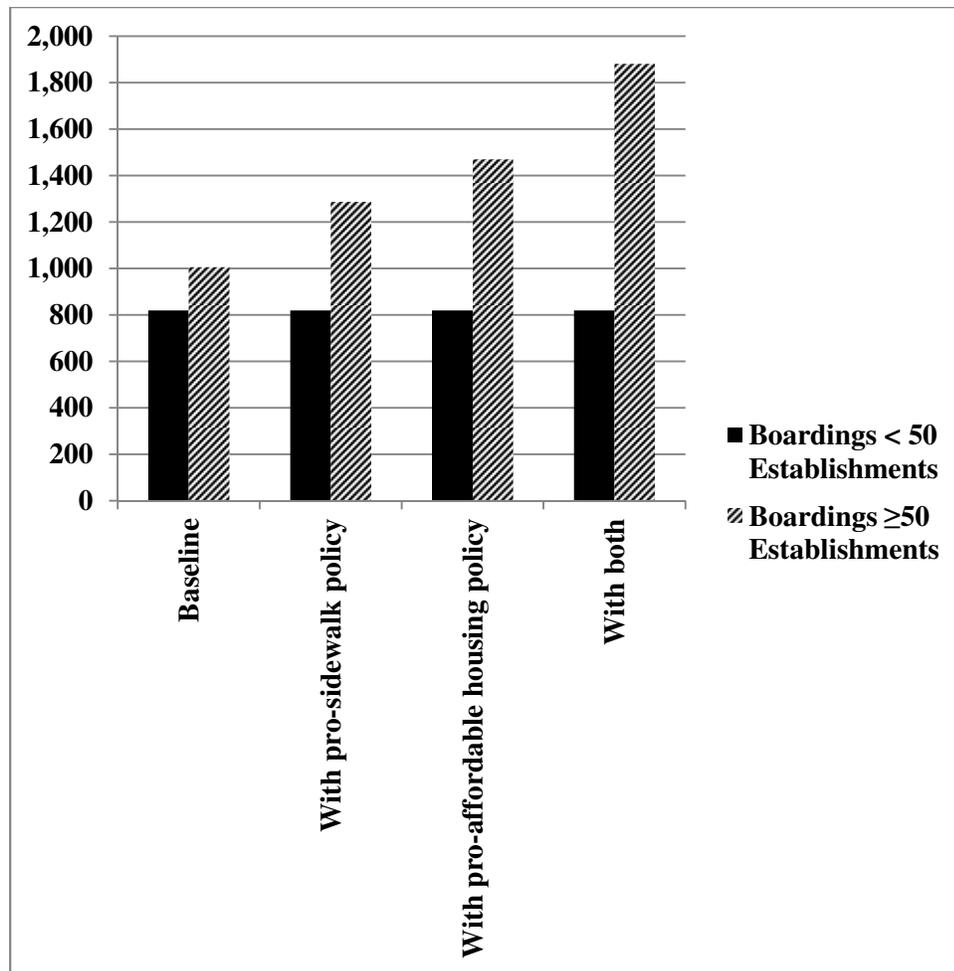
Relationship with Ridership	Strongest	HRT (binary)***	1.434	***
		Number of lines (count)***	1.280	***
		% foreign born residents***	1.179	***
		Pro-affordable housing policy (binary)***	1.144	***
	↑	Distance from next station (kM)*	1.121	*
		% households where workers < cars***	1.116	***
	+	Number of establishments (count)**	1.097	**
		Terminal station (binary)**	1.076	**
		Pro-sidewalk policy (binary)*	1.057	*
		Legacy system (binary)	1.049	
		Median income (\$1K)	1.011	
		% population ≥ 65 yr.	0.971	
		Park and ride (binary)	0.967	
		CR (binary)	0.966	
	Pro-density policy (binary)	0.953		
—	Residential density (housing units/acre)	0.951		
↓	% minority population	0.949		
	% 1-person households**	0.919	**	
	% population < 18 yr.***	0.818	***	
	One-direction station (binary)***	0.725	***	
	Distance from CBD (kM)***	0.698	***	
Strongest	BRT (binary)***	0.659	***	

443 **Model Predictions**

444 As a hypothetical example of the implications of the model results, the following section
 445 presents predicted boardings for archetypal stations in the data. Computing marginal effects for
 446 different transit-supportive policies while holding other variables constant can illustrate the real-
 447 world impacts of the results. Figure 2 shows model predictions for two typical light rail stations,
 448 one with less than 50 surrounding establishments, one with at least fifty. Other than this
 449 difference, and the policy variables being manipulated, these stations are absolutely typical of the
 450 sample: other variables are held at their median/modal values

451 In every case, the typical station with less than 50 surrounding establishments attracts 819
 452 average weekday boardings. No policy variable are significant in the relevant model: it predicts
 453 no change in ridership based on the presence of transit-supportive policies.

454 With no supportive policies, the typical station with at least 50 surrounding establishments
 455 attracts 1,006 average weekday boardings. Predicted boardings increase to 1,287 with an explicit,
 456 pro-sidewalk policy and to 1,471 with a station area-focused pro-affordable housing policy. The
 457 presence of both policies increases the predicted number of boardings to 1,881, nearly double the
 458 predicted boardings with neither policy and more than double the boardings predicted for the
 459 station with less than 50 surrounding establishments.



460

461 **FIGURE 2 Model predictions**462 **CONCLUSIONS**

463 Under the right circumstances, transit-supportive public policies matter. While the actual built
 464 environment is crucial, the models show that policies supporting affordable housing and
 465 pedestrian infrastructure consistently increase transitway ridership *if* they are applied in a station
 466 areas with a sufficient level of activity. The dramatic difference in results upon splitting the final
 467 models by the number of surrounding establishments illustrates the strength of the interaction
 468 between policies and station area activity.

469 It is particularly important to note the strength of the positive relationship between station
 470 area-focused pro-affordable housing policies. This suggests something of a departure from
 471 common planning practice, at least as far as planning for increased ridership goes. Consideration
 472 of transit access in siting affordable housing is not uncommon, but the rationale for it tends to be
 473 one of social equity—an attempt to improve the lives of the specific individuals who will live in
 474 such housing. This research shows that supporting the provision of affordable housing in station
 475 areas enhances transit system efficiency as well. This is an important finding to bear in mind in
 476 the station area planning process, especially considering that affordable housing can be
 477 significantly more fraught with controversy than aspects such as pedestrian infrastructure.

478 The link between surrounding area destinations and pro-sidewalk policies appears fairly
 479 straightforward: policies promoting sidewalks matter if (and *only* if) there are actually things to

480 walk to. The link between pro-affordable housing policies and nearby destinations may, at first,
481 seem less clear. On further consideration, one might contend that station area focused pro-
482 affordable housing policies play a significant role in encouraging transitway usage *given*
483 sufficient opportunity to accomplish one's general daily needs in the station area, as well as
484 affordably live there. Put simply, no matter what level of regional mobility or employment
485 access living near a transitway station offers, if one cannot reach a grocery store, daycare center,
486 etc. without a car, one is likely to own one, regardless of the financial strain that may come with
487 it. Such a conclusion would also follow from the finding of a strong positive relationship
488 between households with fewer cars than workers and transitway ridership.

489 The results recommend station-area focused affordable housing promotion and explicit
490 pro-sidewalk policies in sufficiently active environments as a way of increasing transitway
491 ridership. This is not to say, however, that such policies are not worth pursuing for station areas
492 with more sparse surroundings. The models do not show that affordable housing- and sidewalk-
493 promoting policies are worthless in areas with relatively few potential destinations nearby, but
494 that they are, by themselves, insufficient in such areas. Planning to increase transitway ridership
495 in lower activity areas should focus both on supportive policies and on development of the
496 neighborhood-scale commercial centers that make them relevant.

497 The modeling results speak to the power individual municipalities have to determine the
498 degree of success transit investments serving them will have. They also speak to the limitations
499 of those policies: the basic, necessary conditions of the built environment for the creation of
500 transit-supportive neighborhoods. The findings lay out an agenda for cities served (or planned to
501 be served) by transitways to both align their municipal policies with the goal of promoting
502 transitway ridership, and to set about coaxing their built environments into the needed
503 neighborhood-scale form as well.

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