

1 **Does Built Environment Matter for Innovation?**

2 A quantitative Study of the Physical Assets of Innovation Districts in the United States

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23

24 **Abstract**

25 A growing number of leaders, policy makers, developers, city influencers, and researchers are
26 working on unleashing the concealed economic potential of cities. “Innovation District” is a recent
27 urban model that has emerged as the result of the U.S. economy’s transformation from traditional
28 industrial economies to the knowledge-intensive. The distinguished built environment
29 characteristics of “Innovation Districts” are density, walkability, land use mix, and enhanced
30 transit accessibility. They are identified as the contributors to the innovation-driven economic
31 prosperity through business clustering, creative economy, and regional economic resiliency. A
32 large body of theoretical and qualitative studies have sought to shed a light on the concept of
33 Innovation District. There is, however, a lack of quantitative evidence that support these studies.
34 This national study seeks to fulfill this gap in the literature and examine the relationship between
35 the built environmental characteristics, such as walkability, transit quality and urban form, and
36 innovation generation at the neighborhood level. We used Multi-level Modeling (MLM) to account
37 for the built environment characteristics at both neighborhood and regional levels. Accounting for
38 economic and socio-demographic cofounding variables, we found that innovative firms tend to
39 locate more in dense, pedestrian-friendly, and transit accessible neighborhoods. We also found
40 that a typical neighborhood in a compact region is more attractive for innovative firms than a
41 sprawling region. Our findings confirm the significance of urban form and its physical features on
42 the notion of Innovation Districts theorized by previous studies.

43

44 **Keywords:** Innovation District, built environment, public transit, urban sprawl, compactness,
45 pedestrian-friendly

46

47 **Introduction**

48 During the final day of 2016 Democratic National Convention Hillary Clinton committed to
49 robust U.S. economy through enhancing innovation, small businesses, and technology during the
50 first 100 days of her possible administration (1). It is evident that the U.S. economy relies more
51 and more on innovation and cities; particularly toward the rise of “innovation districts”.
52 Innovation districts are compact, mixed-use, walkable, and transit accessible districts with
53 leading-edge educational institutions and talented human capital and, as a result, are where the
54 innovation generation is accelerated (2). The California’s Silicon Valley and Boston’s Route 128
55 Technology Corridor are successful examples of how built environment could enhance
56 innovation in the urban settings.

57 The emergence of innovation districts is a result of transformation of traditional economy to the
58 modern knowledge economy. The traditional economy used to rely on factories, mass
59 manufacturers and an intensive labor force while the creative economy relies heavily on the
60 entrepreneurs, educational institutions, innovative businesses, and talented human capital (3).
61 This transformation, at least in theory and to some extent in practice, has made significant
62 implications and changes in the structure of urban areas (4) The notion of “industrial districts”
63 where the goal was to provide working and living places for the greater number of workers has
64 changed to the notion of innovation districts where the goal is to provide quality work and life
65 experience for the creative class (5).

66 Several researchers sought to understand the characteristics of innovation districts (2, 6, 7, 3).
67 Some studies focused on the built environmental preferences of the creative class as talented
68 human capital (7, 8). Others applied foundational urban economic theories which focuses on the
69 business clustering and business location decision making (9, 10). The majority of these studies
70 are theoretical and qualitative in nature. There is a lack of quantitative, data-based evidence that
71 support these theories and show how built environment at both neighborhood and regional scales
72 influence innovation generation in an urban setting.

73 This national study aims to address this gap by examining the relationship between built
74 environment characteristics and innovation generation. Using Multi-level Linear Modeling
75 (HLM), we account for the built environment characteristics at both neighborhood and regional
76 levels. The major built environment characteristics at the neighborhood level are the degree of
77 mixed-use, walkability, transit availability/quality, and street accessibility. We account for urban
78 form at the regional scale through the degree of urban sprawl as a proxy for regional
79 accessibility.

80 This study employs disaggregated data of innovation generation from the Small Business
81 Innovation Research (SBIR) program coordinated by the Small Business Administration (SBA).
82 The SBIR program has been awarding the competitive federal grants to the innovative small
83 businesses since 1982 (11). Our study covers a Fifteen-year timeframe and includes all the SBIR
84 awarded firms from 2000 to 2015. One great advantage of using SBIR over other databases is the
85 fact that it is designed to support small businesses which otherwise could not compete with large
86 firms such as Google and Apple. We gathered neighborhood level data from various data sources
87 and we used the widely cited Ewing and Hamidi (2014) metropolitan compactness index for the
88 measure of urban compactness.

89 **The Rise of Innovation Districts**

90 The emergence of innovation districts is the result of several leading and historical trends. First,
91 business clustering and geographic proximity play a significant role. Jacobs (1970), and more
92 recently Katz and Bradley (2013) describe the necessity of spatial proximity for industries,
93 businesses and universities in order to have access to human capital and flow of knowledge with
94 lesser marginal costs (12, 4).

95 The other significant contributor to the emergence of innovation districts is the change in the
96 American family structure and preferences. Traditionally, the isolation of work and life was the
97 preference of a typical American household living in suburban areas. Residential preferences has
98 changed substantially in recent years as a result of socio-demographic shifts such as delayed
99 marriage, fewer number of children in the family and a smaller household size which has led to
100 an increasing preference for living in compact and walkable neighborhoods (13). Previous
101 studies show an increasing trend in the number of people who live within three miles of the
102 central business district (14, 15). The preference for walkable distance to restaurants, retail,
103 cultural, and educational institutions is even greater for the creative and educated class (8).
104 Walkability is also recognized as an important factor for college-educated millennials when they
105 choose a place to live and work after graduation (15).

106 An educated working class and creative human capital are the generators of knowledge
107 economies (7). High-tech firms and businesses follow the knowledge workers and so they tend to
108 relocate to denser, more compact and accessible urban areas (16). Therefore, the areas that offer
109 appealing living environment for them, are more successful in attracting the knowledge intensive
110 companies which in turn results in higher innovative activities and more economic prosperity.
111 Detroit MI, Houston TX, and Buffalo NY are examples of cities have sought to incorporate the
112 concept of innovation district in their future plans (17, 4). Some regions such as Boston have
113 already started to implement the concept of innovation through the Boston's Route 128
114 Technology Corridor (18).

115 **Characteristics of the Innovation Districts**

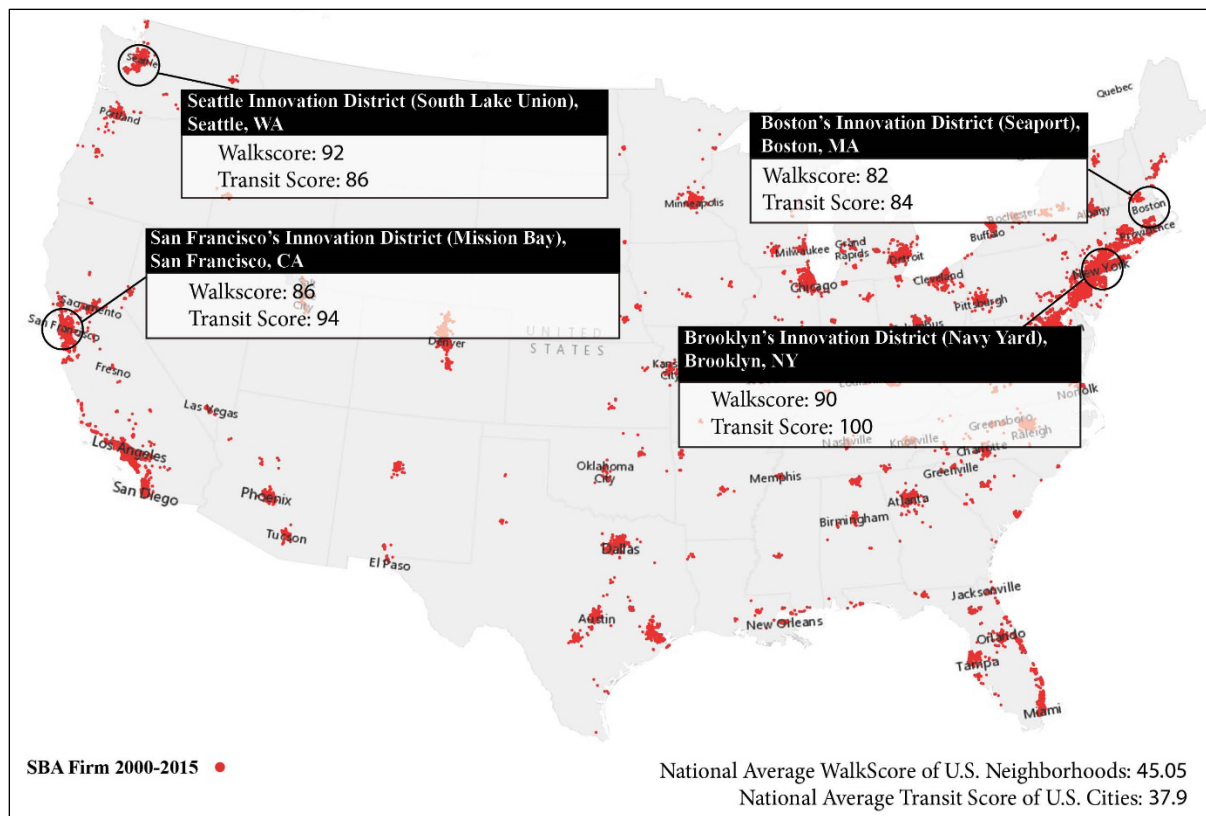
116 Innovation districts are compact, transit-accessible areas that provide a mix of housing, offices,
117 and retail. Innovation districts bring together leading-edge research institutions and high-tech
118 companies and connect them to start-ups, business incubators, and accelerators (2). As a result,
119 innovation districts facilitate innovation generation through creating, attracting, and retaining the
120 creative class and knowledge-intensive companies (3, 19). So the question is why some places
121 support innovation generation more than others. What locational factors contribute to more
122 innovation in these places?

123 In 2014 the Brookings Institution released a comprehensive, and the first of its-kind report, on
124 the Innovation Districts (2). According to this report, Innovation Districts have three categories
125 of assets: economic assets, networking assets and physical assets.

126 Economic assets are referred to as the presence of 3 general business categories. The first
127 category includes the innovation drivers functioning as universities, anchor institutions, high tech
128 research-oriented businesses (such as material science, energy technology and nanotechnology),
129 creative firms (like art-tech companies), and small artisan manufacturing or textile production.
130 The second category presents to support the growth and innovative activity of the first category

131 serving as incubators, start-ups, accelerates, etc. The third category includes the businesses that
132 build the neighborhood sense of an innovation district like cafes, restaurants, grocery stores,
133 bars, local retails and hotels. The economic assets emphasize the importance of spatial proximity
134 and clustering of economic activities.

135 Networking assets concentrate on the long been acknowledged value of networking through
136 competition, collaboration, and learning between companies (20). These assets consist of strong
137 ties and weak ties. The strong ties enhance the contact and communication between actors of
138 similar fields and by this nature are workshops, training sessions, conferences, and industry
139 specific blogs for specific fields. The weak ties tend to either build new or enhance existing
140 cross-sector relationships. Weak ties emerge in form of innovation centers, city halls,
141 hackathons, and the training sessions that aim to help businesses grow. The networking assets
142 emphasize the importance of social interaction, which provides the chance of serendipitous
143 encounters, networking, and knowledge spillover.



144
145 **FIGURE 1** Selective Physical Assets of Four Innovation Districts and Their Comparison to the
146 National Average (Source: Authors)

147
148 The innovation districts' physical assets enhance the innovative ecosystem through providing
149 density, high level of mix use and street connectivity. They include three categories of built
150 environmental characteristics. The first category belongs to the public realm and includes the
151 neighborhood-scale public spaces such as neighborhood parks, exhibitions, concert areas, plazas
152 and public realm's eateries which will strengthen the social interactions. To get the most and best

153 out of quality public spaces, they need to be integrated within a dense and highly mixed-use
154 development.

155 The second category has to do with the preferences of the innovation demographics. They prefer
156 the micro/mixed income housings built in the mixed-use and dense complexes of houses, offices,
157 co-working areas and stores. Connectivity of the public realm's and private realm's physical
158 assets is considered within the third category. The third category is the walkable and bikable
159 street network with the enhanced street crossings. By the same notion, high quality and
160 accessible public transit serves both internal and regional connectivity. Boston's Silver Line,
161 Houston's Red Line, and the future Detroit's M-1 line are the transportation infrastructure
162 contributors to the development of innovation districts in each of these cities.

163 **Regional Accessibility and Innovation**

164 Proximity, or in a broader term "accessibility," of industries, business services, and anchor
165 institutions is a driving force for innovation (21, 22, 23, 24, 25). Accessibility provides
166 networking opportunities for these entities through raising interactions, collaborations, and
167 knowledge spillover (26). Compact areas with stronger transportation accessibility foster
168 networking opportunities between people via enhancement of social capital (27). In contrast,
169 urban sprawl could provide less opportunity for knowledge flow and interaction between
170 innovative people, which in turn could result in lesser innovation and knowledge production.

171 Metropolitan Sprawl is characterized in the literature as areas with poor accessibility. In scattered
172 or leapfrog development, residents and service providers must pass vacant land on their way
173 from one developed use to another. In commercial strip development, the consumer must pass
174 other uses on the way from one store to the next — the antithesis of multipurpose travel to an
175 activity center. Of course, in low-density and single-use development, everything is far apart
176 because of large, private land holdings and segregation of land uses. In sprawl, poor accessibility
177 of land uses to one another may leave residents with no alternative to miles and miles of
178 automobile travel. So the question is how poor accessibility might affect regional innovative
179 capacity.

180 There is little evidence in the literature on how built environmental attributes and the resulting
181 accessibility influence innovation generation at different geographic scales. This study seeks to
182 address these gaps by employing multilevel modeling and gathering data from various sources at
183 the neighborhood and regional levels. In the next section, we will explain data, variables and the
184 methodology.

185 **Methods**

186 **Data and Variables**

187 The complete set of variables and data sources are shown in Table 1. Our dependent variable is
188 the number of innovative businesses in each census tract are obtained from SBA. The SBIR
189 database is one of the most widely used measures of innovation capacity under the innovation
190 counts category (28, 29). SBA serves as the coordinating agency for two national innovative
191 award programs: SBIR and Small Business Technology Transfer (STTR). Both recognize
192 innovative activities accomplished by small businesses. SBIR awards are given to innovations
193 developed solely by small businesses, while STTR awards are given to collaborative innovative

194 activities between small businesses and research institutions. SBIR and STTR awards are granted
 195 in three phases.

- 196 • Phase I recognizes innovative concepts from small businesses. At this phase, the
 197 feasibility of idea and proof of concept are assessed.
- 198 • Phase II covers the full research and development for a prototype among Phase I
 199 grantees.
- 200 • Phase III is the commercialization phase. Phase I/II grantee products, services,
 201 technologies, or processes are produced and delivered (11).

202 In this study, we use the SBIR and STTR database as our measure of innovation capacity. In
 203 comparison to other measures, the SBIR and STTR database is inclusive and does not favor large
 204 companies as the R&D database does. Also, it evaluates and recognizes innovation for both
 205 process and product through a competitive three-phase assessment method from proof of concept
 206 to development and commercialization. Finally, the chance of overestimation of innovation is
 207 less than other approaches, and it introduces the more valuable innovations market wide.

208 Our study includes both SBIR and STTR phase II between 2000 and 2015. SBA provides a
 209 substantial award in phase II to research and development of an innovation. The fifteen year
 210 time frame allows us to obtain a consistent measure of innovative firms than relying on single-
 211 year data. We used walk score as the measure of walkability at the neighborhood level.

212 TABLE 1 Data and Variables

Variables		Data Sources
Level 1 Dependent variable		
<i>SBA_Firms</i>	Number of SBIR/STTR Phase II firms between 2000-2015	SBA 2016
Level 1 Independent variables (census tract level)		
<i>pctWrk</i>	Percentage of working age population for 2010	Census 2010
<i>htechcap</i>	Average number of high tech jobs per capita	LED 2011
<i>WlkScr</i>	Census tract level Walk Score	Walk Score Inc. (2010)
<i>Transfreq</i>	Average transit service frequency	SLD-EPA (2010)
<i>Entropy</i>	Land use mix within a census tract (entropy index based on net acreage in different land use categories that ranges from 0, where all developed land is in one use, to 1, where developed land is evenly divided among uses)	Computed in the study using 2010 LEHD data
<i>Actden</i>	Activity density (sum of population and employment divided by gross land area in square miles)	Computed in the study using 2010 LEHD and census data
<i>intden</i>	Intersection density	Computed using TomTom 2007 database
<i>hincome</i>	Average household income	ACS 2007-2011
<i>Pop000</i>	Census tract population in 1000s	Census 2010
Level 2 Independent Variables (metropolitan level)		
<i>MSAindex</i>	metropolitan compactness index for 2010	Ewing and Hamidi (2014)
<i>Crime</i>	violent crime rate per 100,000 population	FBI Uniform Crime Statistics 2010
<i>UniR&D</i>	The university R&D expenditure 2011	NSF

213 Our first built environment variable is census tract level Walk Score. It was computed using data
214 from Walk Score, Inc. to measure proximity to amenities, with different amenities weighted
215 differently and amenities discounted as the distance to them increases up to one mile and a half,
216 where they are assumed to be no longer accessible on foot.¹

217 We accounted for transit availability and quality through the transit service frequency data
218 retrieved from Smart Location Database (SLD) developed by Environmental Protection Agency
219 (EPA). Transit service frequency dataset is calculated based on the average number of scheduled
220 transit service per hour during the evening peak period (4:00-7:00 pm on a weekday) for all
221 block groups in the U.S. we aggregated up the block group data and computed this variable at the
222 census tract level.

223 Other neighborhood level built environmental variables used in our model are activity density,
224 land-use mix and intersection density. Intersections are where street connections are made and
225 cars must stop to allow pedestrians to cross. The higher the intersection density, the more
226 walkable the city (30). Intersection density has become the most common metric in studies of
227 built environmental impacts on individual travel behavior (31). Our land use mix variable, an
228 entropy measure, equals 1 for block groups with equal numbers of jobs in each sector; 0 for
229 block groups with all jobs in a single sector within the ring; and intermediate values for
230 intermediate cases. The sectors considered in this case were retail, entertainment, health,
231 education, and personal services.

232 At the regional level, we used the metropolitan compactness index developed by Hamidi and
233 Ewing (2015). The metropolitan compactness indices consist of 21 built environment variables
234 of four distinct dimensions: development density, land-use mix, population and employment
235 centering, and street connectivity. The indices were constructed so that the more compact a
236 metropolitan area was, the larger its index value.

237 At the MSA level, we also controlled for the R&D of high-tech industries, university R&D
238 expenditure, and crime rate. We employed the 2011 Longitudinal Employer-Household
239 Dynamics (LEHD) from the U.S. Census Bureau, which is based on North American Industry
240 Classification System (NAICS) codes for the R&D of high-tech sectors. The R&D of high-tech
241 companies is computed as the proportion of high-tech jobs to population. Suggested by the
242 literature, we included two high-tech sectors 1) mining, quarrying, and oil and gas extraction and
243 2) manufacturing industries. We used Standard Industrial Classification (SIC) codes of 28, 35,
244 and 36 as high-tech employment sectors (32) and converted them into NAICS.

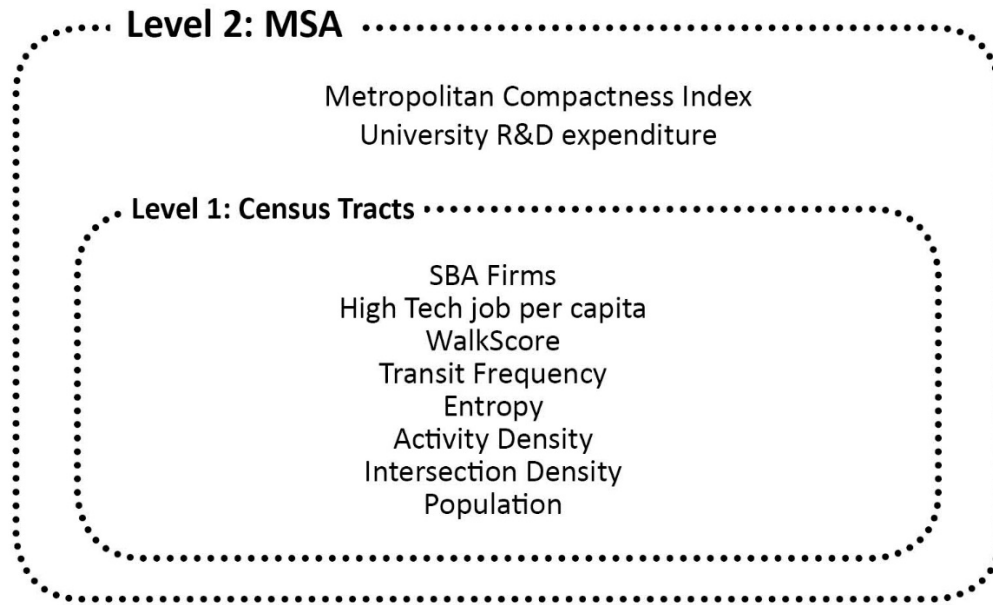
245 As for the university R&D expenditure, we gathered data from the publicly available 2011
246 National Science Foundation Business Research and Development and Innovation Survey. Using
247 Geographic Information System (GIS), we aggregated and computed the university R&D
248 expenditure at the MSA level.

249 [Analytical model](#)

250 As shown in table 1, the data used in this analysis have a “nested” structure and should be
251 analyzed accordingly. Since the census tracts located in an MSA share the characteristics of that

¹ A grocery store, for example, gets three times the weight of a book score. The distance decay function starts with a value of 100 and decays to 75 percent at a half mile, 12.5 percent at one mile, and zero at 1.5 miles.

252 MSA, such as the metropolitan compactness index, and could not be treated as independent of
253 each other. The nesting feature causes the dependence among cases, violating the independence
254 assumption of OLS regression. In this situation, the standard errors of regression coefficients
255 associated with MSA characteristics based on OLS will then be underestimated, and regression
256 coefficient themselves will be inefficient (33).



257
258 FIGURE 2 Conceptual Framework Showing the Nesting Structure of Variables

259
260 Multilevel Modeling (MLM) addresses the issue of nesting structure and dependence among
261 cases and leads to more accurate coefficient and standard error estimates. In this analysis, the
262 numbers of innovative firm were regressed on neighborhood characteristics in level-1 models.
263 The intercepts and coefficients of level-1 models were regressed on regional characteristics in
264 level-2 models. Initially, we estimated two different models. In the first model, only the intercept
265 was allowed to randomly vary across respondents, while all of the regression coefficients were
266 treated as fixed. These are referred to as “random intercept” models. Later on, regression
267 coefficients were allowed to randomly vary across higher level units as well, and interactions
268 between levels were allowed. These are called “random coefficient” models. As cross level
269 interaction terms seldom proved significant, we reverted to the random intercept model. Only
270 this model are presented in the next section.

271 The other statistical complication relates to our dependent variable. Our dependent variable is the
272 number of innovative firms (count data) in a census tract which is a count. Two basic methods of
273 analysis are available when the dependent variable is a count, with nonnegative integer values,
274 many small values and few large ones. The methods are Poisson regression and negative
275 binomial regression. The two models differ in their assumptions about the distribution of the
276 dependent variable. Poisson regression is appropriate when the dependent variable is
277 equidispersed, meaning the variance of counts is equal to the mean. Negative binomial
278 regression is appropriate when the dependent variable is overdispersed, meaning that the
279 variance of counts is greater than the mean. Popular indicators of overdispersion are the Pearson

280 and χ^2 statistics divided by the degrees of freedom, so-called dispersion statistics. If these
 281 statistics are greater than 1.0, a model is said to be overdispersed (34). By these measures, we
 282 have overdispersion of innovative firm counts in our dataset, and the negative binomial model is
 283 more appropriate than the Poisson model. Therefore we employed negative binomial multi-level
 284 modeling in this study.

285 Results and Discussion

286 We estimated a multi-level negative binomial model using HLM 6.8. The results of the best
 287 fitted model is presented in Table 2. The coefficients of most variables are significant and have
 288 the expected signs.

289 Our findings support the notion of “innovation districts”. According to the literature, the
 290 innovation districts thrive where housing, restaurants, retails, anchor institutions and small-scale
 291 innovative activities are collocated near transit (2). We find that the number of innovative firms
 292 in a census tract is positively and significantly associated with walkScore, transit frequency and
 293 activity density although in case of activity density it approaches the significance level at 0.1.
 294 The most significant variable in the model is WalkScore which confirms the importance of
 295 walkability on innovation generation. Also, neighborhoods that enjoy access to a quality transit
 296 service are more attractive for innovative small firms.

297 Our results confirm that innovative firms are more likely to locate in mixed-use neighborhoods
 298 where there is a mix of retail, entertainment, health, education and other major destinations are,
 299 although the relationship is not significant. We found a significant negative relationship between
 300 the intersection density and the number of innovative firms in a census tract. More intersections
 301 mean more stops and slower traffic flow (35) which could be unpleasant to firms, innovative
 302 businesses and commuting employees.

303 TABLE 2 Negative Binomial Reression Analysis of Innovation Generation, Buit Environment and
 304 Transpotaion Accessibility for Urban Cesus Tracts

	Coeff.	Std. error	t-ratio	p-value
<i>Constant</i>	-0.233	0.455	-0.512	0.610
<i>htechcap</i>	0.838	0.234	3.581	0.001
<i>WlkScr</i>	0.00964	0.0016	6.116	0.000
<i>Transfreq</i>	0.000553	0.000225	2.464	0.014
<i>Entropy</i>	0.033	0.172	0.191	0.848
<i>Actden</i>	0.000004	0.000002	1.690	0.091
<i>intden</i>	-0.0053	0.001	-5.166	<0.001
<i>Pop000</i>	0.00555	.0041	1.353	0.176
<i>MSAindex</i>	0.0065	0.0024	2.688	0.009
<i>lnuniR&D</i>	.026	.0275	0.942	0.350
<i>Pseudo-R²</i>	0.517			

305
 306 At the MSA level, we found that the number of innovative firms in a census tract is positively
 307 and significantly associated with the MSA level of compactness. In other words, a typical census
 308 tract in a compact metropolitan area is more attractive for innovative firms than the same census
 309 tract in a sprawling metropolitan area. This result squares with our expectations based on
 310 previous studies and evidence (7, 21, 25, 23). Sprawling areas are heavily car dependent, do not

311 accommodate regional accessibility and reduce opportunities for social life which could inspire
312 the creative class to socialize and exchange ideas (7, 4, 2, 3). Compact places increase the chance
313 of serendipitous encounters, networking, and knowledge spillover (36, 2).

314 Our model suggests a positive but insignificant relationship between the university R&D
315 expenditure at the regional scale and the number of neighborhood level innovative firms. Less
316 than 10 percent of SBA awards between 2000 and 2015 were allocated to the university-
317 businesses collaboration innovation. Thus, the collaboration between high-tech industries,
318 particularly small businesses, and universities needs further studies.

319 All our control variables have the positive sign in this model, but they are not always significant.
320 The size of a census tract does not significantly affect its likelihood of having higher number of
321 innovative firms. Number of high-tech firms in a neighborhood is positively and significantly
322 associated with the number of innovative firms. This agrees with the theory of spatial proximity
323 of high-tech firms as the main contributor to the innovation. The industry R&D and regional
324 crime rate are the other two insignificant variables.

325 The result of this study represents the significance of the three categories of physical assets as the
326 key components the innovation district (2). Both public realm and privately built physical assets
327 result in the development of an innovative ecosystem when are highly land use mix and dense.
328 The innovative ecosystems at the neighborhood level will generate more innovation when they
329 are not isolated. Our findings confirm the significance of both enhanced internal connectivity
330 through walkability, and an effective external connectivity to broader opportunities in the region
331 through public transit for creating an innovative ecosystem that supports innovation generation.

332 Furthermore, the creative class would prefer to live in places that are dense and mixed-use and
333 offer opportunities for walking, biking, and transit use. The creative class is the driver for
334 regional innovation and places that offer appealing living environment for them, are more
335 successful in attracting them and knowledge intensive companies which, in turn, results in higher
336 innovative activities and more economic prosperity.

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