

Influence of Population Concentration in Urban Agglomeration on Corporate Green Innovation

Author's contribution

The sole author designed, analyzed, interpreted and prepared the manuscript.

ABSTRACT

Green innovation is considered to be able to coordinate economic development with ecological environment improvement. Scholars are increasingly focusing on the factors influencing corporate green innovation capability. The analysis is based on the panel data econometric model with fixed effects to examine the relationship between population concentration in urban agglomeration and corporate green innovation. By analyzing company-level data of Chinese A-share listed companies from 2009 to 2019, it was discovered that: (i) Population concentration in urban agglomerations has a significant positive impact on corporate green innovation in the domain. (ii) This relation is stronger for firms in central and western regions, low marketization regions and small cities. (iii) This relation is stronger for small and medium-sized and state-owned enterprises. The finding will provide new insights into the relationship between population concentration in urban agglomeration and corporate green innovation in the domain, which will provide empirical evidence for exploring new urbanization in China.

Keywords: Urban agglomeration; green innovation; regional integration development; corporate sustainable growth.

1. INTRODUCTION

Green innovation is becoming the key to the new round of global industrial revolution and technological competition. As a kind of excellent instrument to balance economic development and environmental protection[1], green innovation is expected to provide inexhaustible driving force toward economic sustainable development and break the bottleneck of economic development constraints [2-4]. For enterprises, green innovation can help them achieve a win-win situation in terms of economic development and environmental protection by helping them to achieve energy saving and emission reduction as well as assisting them to produce green and differentiated products. As the country with the highest total carbon emissions, China is under tremendous pressure to reduce emissions. In order to promote the development of low-carbon transformation, China had proposed the strategic goal of achieving peak carbon by 2030 and carbon neutrality by 2060. Green innovation is the key to energy saving and emission reduction, which is an effective method to effectively solve environmental pollution [5]. Enterprises are the important subjects to implement the strategic goals of carbon peak and carbon neutral. By accelerating the development of quantity and quality of green technology innovation, enterprises can better realize the prevention and control of environmental pollution, promote the green and low-carbon transformation of enterprises and green development of economy and society, and then continuously reduce carbon emissions, which can provide important technical support to achieve the goal of carbon peaking and carbon neutrality.

However, in the context of China's ecological civilization practice and high-quality development, green innovation is still at a low level in China. The main reason for this phenomenon is the lack of green innovation exchange platform in China and the low level of inter-regional green innovation cooperation [6]. Another factor is the general situation of domestic regional segmentation in China [7].

40 New urbanization is an effective way to alleviate the situation of domestic regional
41 segmentation. The 20th National Congress of the Communist Party of China pointed out that
42 China should build a coordinated development pattern of small, medium and large cities
43 based on urban clusters and then promote a new type of people oriented urbanization. The
44 concentration of industry and population in regions with economic advantage is an
45 inexorable long-term trend of global economic development. In 2022, Chinese government
46 issued a policy to support the integration of counties into the construction and development
47 of neighboring large cities by pooling resources, promoting growth, stimulating innovation,
48 optimizing the division of labor, and promoting competition through unified large market.
49 Gao, Zhang [8] proposed that the core of industrial and population clustering is to enable
50 neighboring regions to achieve regional comprehensive competitiveness. Modern economic
51 theory suggests that population and industry concentration can lead to agglomeration effects
52 and large-scale economy can promote economic growth and green innovation through the
53 mechanism of sharing, matching and learning. Labor force is the most important and mobile
54 resource element for regional economic and social development. My research explores the
55 impact of population concentration in urban agglomeration on regional green innovation
56 vitality by studying corporate green innovation activities.

57 Many empirical studies focusing on European and American countries have found that urban
58 agglomerations play an important role in regional economic development [9, 10]. By
59 comparing the urban systems of China, the United States, and European countries, Glaeser,
60 Ponzetto [11] find that the urban system in the United States is mainly a few metropolitan
61 areas dominated by large central cities, while the urban system in Europe is the network of
62 cities formed by small and medium-sized cities. Relatively speaking, China is more suitable
63 for the urban network model because of its huge population size. The increase in population
64 concentration in urban agglomeration can promote urban economic development by
65 optimizing industrial structure and promoting regional integration. In terms of the internal
66 structure of urban agglomeration, the development of the central cities and their radiation
67 capacity play an important role in the overall development of urban agglomeration, and the
68 interconnectedness of non-central cities also contributes to the realization of this externality.
69 Alonso [12] proposed that, by taking advantage of the surrounding areas, a city can get more
70 development opportunities and perform better in terms of economic. And this idea has been
71 further refined and tested by related scholars[9, 13]. From the perspective of coordinated
72 development, the development of urban agglomeration can help to reduce the financial
73 development gap among regions. As the division of functions within the urban agglomeration
74 continues to advance, the regional economic gap will also gradually decrease. From the
75 perspective of spatial externality, agglomeration can reduce the cost of flow of factors, goods
76 and information, which means that economic agents can enjoy larger labor and demand
77 market and have access to more technology and information [14, 15]. Externalities arising
78 from upstream and downstream industry linkages and technological knowledge spillovers
79 can improve economic performance, which is empirically reflected in productivity gains and
80 labor wage premiums [16, 17]. All the characteristics of population concentration of urban
81 agglomeration mentioned above are conducive to the corporate level of green innovation.

82 In the context of new urbanization, population concentration in urban agglomeration can
83 provide high-quality human capital to support corporate green innovation activities. The
84 adjustment of the regional economic development model is likely to profoundly affect
85 corporate green innovation decisions. Accordingly, urban agglomerations are one of the
86 important spatial carriers of China's economic development in the future. The clustering of
87 green innovation factors in urban agglomerations is important for achieving the goal of
88 carbon neutrality and carbon peaking by promoting the coordination of economic
89 development and environmental protection. This paper investigates the relationship between
90 population concentration in urban agglomerations and corporate green innovation in the
91 domain, hoping to provide micro-level evidence for the new people-oriented urbanization
92 strategy proposed in China in recent years.

93 Based on the theoretical framework of new economic geography, taking Chinese
94 development history of urbanization into account, my study examines whether the population
95 concentration in urban agglomeration is conducive to green innovation of firms. Using
96 Chinese A-share listed companies from 2009-2019 as the sample, my study constructs
97 indicator based on the geographical location characteristics of the companies to empirically
98 test the relationship between the population concentration in urban agglomeration and
99 corporate green innovation. The result shows that with the increase of population
100 concentration in urban agglomeration, the level of corporate green innovation tends to
101 increase significantly, which holds for both large and small cities. On average, when the level
102 of population concentration in urban agglomeration rises by one standard deviation, the
103 companies green innovation will on average increase by 0.06 percentage points. From the
104 perspective of geographic location, compared with eastern cities, firms in central and
105 western cities benefit more from the population concentration in urban agglomeration.
106 Compared with other existing studies, my study contributes to the literature in three ways.
107 Firstly, enrich the literature on corporate green innovation. Although scholars have paid
108 increasing attention to corporate green innovation, there is relatively little literature
109 examining the impact of population concentration in urban agglomeration on corporate green
110 innovation. Secondly, my study examines the economic impact of population concentration
111 in urban agglomeration at the micro level. Most of the existing studies on urban
112 agglomeration have focused on the city level, which pays more attention to the overall
113 economic development of urban agglomeration. There are relatively few empirical studies
114 on the population concentration in urban agglomeration and how it affects corporate
115 behavior. Thirdly, my study can provide evidence to support the national decision of
116 choosing population clustering in cross-regional urban agglomeration as a focus for future
117 regional development.
118 The remainder of the paper proceeds as follows. In Section 2, I sort out the relevant
119 literature. In Section 3, I present the sample selection and the construction method of
120 variables. In Section 4, I give empirical results and explanations. Finally, Section 5 is the
121 conclusion and discussion of this paper.

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2. LITERATURE REVIEW

125 Recently, there are many studies paying attention to the extension and influencing factors of
126 green innovation. Green innovation refers to the innovation of environment and ecology [18],
127 which is defined as the hardware and software innovation related to green processes and
128 products, including areas such as energy conservation, pollution prevention, waste recycling,
129 green product design, enterprise environmental management and so on [19].Betz [20] point
130 out that technological breakthrough can bring great benefit to the development of enterprise
131 and society.
132 The existing literature studies green innovation mainly at two levels: one is the meso- or
133 macro-level, and the other is the firm level. Some scholars study the topic of green
134 innovation from the perspective of government policy. Xiang, Liu [21] and Xiang, Liu [21]
135 point out that for state-owned enterprises, political connections have a significant positive
136 impact on enterprises' green innovation by promoting investment in organizational capital
137 and R&D. Additionally, compared to the debt and equity financing, government subsidies
138 have an even greater impact on green innovation of firm. Chen, Zhu [22] examine the
139 relationship between local debt and corporate green innovation and find that local
140 government debt is significantly and negatively related to corporate green innovation. Zhang,
141 Liang [23] study the relationship between environmental regulation and green innovation and
142 suggest that environmental laws in China have an important influence on the process of
143 green innovation. Pan, Cheng [24] study the relationship between environmental policy and
144 green innovation in the sample of regions with high pollution level, drawing a conclusion that
145 green innovation is not strongly influenced by environmental regulations in regions with high

146 pollution level. Moreover, some studies show that the green innovation can be affected by
147 many other meso and macro factors such as media coverage[25], capital market opening[26],
148 the construction of innovative cities [27] and so on.

149 In addition, some scholars discuss the topic related to green innovation from the perspective
150 of corporate behavior. Amore and Bennesen [28] point out that corporate governance has
151 significant implications for corporate green innovation performance. This is demonstrated by
152 the fact that ineffective corporate governance has a negative impact on green innovation
153 activities. Amore and Bennesen [28] further find that personnel in the R&D department and
154 the expenditure on R&D promote corporate green innovation. Qi, Zeng [29] indicate that
155 foreign customers have a significant positive effect on the use of green product innovation
156 strategies and processes by firms. Bin Yousaf, Ullah [30] study the relationship between
157 board traits and innovation activities, drawing a conclusion that board capital has a
158 significant positive impact on green innovation. Moreover, some studies show that the green
159 innovation can be affected by many other corporate factors such as overseas experience of
160 executives [31], academic experience of senior management[32], CEO arrogance [33],
161 female board members[34], corporate resources[21] and so on.

162 A few scholars have also discussed the impact of external shocks such as terrorism and
163 natural disaster on green innovation. For example, using panel data of OECD countries from
164 1975 to 2018, Zheng, Feng [35] examine the impact of natural disaster on green innovation
165 performance, concluding that the occurrence of natural disaster hinders green innovation
166 activities. In addition, Zheng, Feng [36] point out that terrorism has a negative impact on the
167 invention of renewable energy technologies in OECD countries.

168 Although various perspectives have been adopted in the literature to explore the factors
169 influencing green innovation, the relationship between the population concentration in urban
170 agglomeration and green innovation has not been fully studied in the previous literature so
171 far. Economic globalization requires the enhancement of urban agglomeration, which has
172 become a very important model of economic development in modern countries. The
173 development of urban agglomeration and its impacts have become a focus of scholarly
174 attention [37-39]. As an important way to drive China's economic development and
175 accelerate urbanization, the construction of urban agglomeration is transcending regional
176 boundaries and becoming "nodes" connecting different regions and even the global
177 economy, which facilitates the achievement of the 14th Five-Year Plan and the 2035 Long-
178 Range Goals. Population concentration in urban agglomerations accelerates the flow of
179 innovation factors between cities. As an organic system composed of multiple spatially
180 adjacent and closely connected cities, urban agglomeration can integrate innovation
181 resources on a regional scale and make up for the lack of innovation competitiveness of
182 individual cities. The construction of urban agglomeration connects the economic activities of
183 different regions into a whole mainly through the networked development of transportation
184 infrastructure, which extends the borders of cities and urban agglomerations. Urban
185 agglomeration breaks the limitation of knowledge in spatial scope, which can promote the
186 generation of innovation, reduce the cost of regional flow of innovation factors, and increase
187 the possibility of economic activity clustering. Therefore, we need to pay more attention to
188 the potential impact of population concentration in urban agglomeration on green innovation,
189 which will not only help China's urbanization level steadily increase, but also help us
190 promote sustainable economic development from the perspective of green innovation.

191

192 3. MATERIAL AND METHODS

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194 My study examines the relationship between population concentration in urban
195 agglomeration and green innovation by using A-share listed companies' data from 2009-
196 2019. I collected financial and ownership data of sample companies from the CSMAR
197 database, excluding financial listed companies, companies with unusual listing status during
198 the sample period and companies with abnormal financial indicators. Finally, there are

199 19953 annual company observations in the sample. The source of the corporate innovative
200 data is the CNRDS database. The city-level data involved are obtained from the China City
201 Statistical Yearbook and the China Urban Construction Statistical Yearbook, and some
202 missing data are supplemented according to the statistical yearbooks and statistical bulletins
203 of each province.

204 **3.1 Empirical methods**

205 Fixed effects models are widely used in the study of panel data, such as the fixed effects
206 models constructed by scholars to study the effect of Academic workstations on corporate
207 green innovation[40] and robot adoption on green innovation[41]. Inspired by previous
208 studies, this study constructs a panel fixed effects model with industry fixed effect and time
209 fixed effect to examine the relationship between population concentration in urban
210 agglomeration and corporate green innovation in the domain. The difference between this
211 model and others is the inclusion of industry and year fixed effects, which can well solve the
212 problem of variation over time and industry. Therefore, the model can objectively reflect the
213 relationship between population concentration in urban agglomeration and corporate green
214 innovation in the domain.
215

$$216 \quad Y_{ijt} = \beta_0 + \beta_1 \text{Concentration}_{jt} + \beta_2 \text{Scale}_{jt} + \text{Controls}_{ijt} + \text{Year} + \text{Ind} + \varepsilon_{ijt} \quad (1)$$

217 Where Y_{ijt} is the proxy variable for the corporate green innovation level of firm i in period t ,
218 which includes LnTotal, LnInva and LnUma. $\text{Concentration}_{jt}$ refers to the population
219 concentration level in urban agglomeration in period t . The control variables selected for this
220 paper include two levels: firm-level and city-level. Firm-level control variables include
221 Cashflow, Lev, ROA, PPE, Top1, Seperation and Age. City-level control variables include
222 Scale, Fiscal_e, Loan, Road, Sec_ind_ratio and Fdi. This study mainly focuses on the
223 coefficient β_1 . If β_1 is significantly less than 0, it means that population concentration in urban
224 agglomeration is negatively related to corporate green innovation, while the opposite is true
225 if it is significantly greater than 0.

226 **3.2. Variables**

227 **3.2.1 Population concentration in urban agglomeration**

228 The explanatory variable of this study is the population concentration in urban
229 agglomeration, which can be interpreted as the summation of urban population sizes within
230 urban agglomeration. Considering that interactions vary with distance, this summation is
231 weighted by the inverse of the inter-urban distance, and the variable is constructed as
232 follows:

$$233 \quad \text{Concentration}_{jt} = \sum_{k \in D, k \neq j} \frac{\text{pop}_{kt}}{d_{kj}} \quad (2)$$

234 Where j represents the city, t represents the time, D is the collection of cities within a certain
235 geographical distance around city j , and pop is the population size of city k during the t
236 period. Urban population size data comes from the China Urban Statistical Yearbook. d is
237 the geographical distance between city k and city j . Our study calculates the distance
238 between cities using the longitude and latitude coordinates of the city government location
239 on Baidu map. Moreover, the preliminary calculation results show that the average distance

240 between a certain city in China and its nearest five cities is 151 kilometers, so our study set
 241 the geographical range of city cluster D to 150 kilometers.

242 **3.2.2 Green innovation**

243 Green innovation is a form of innovation that is guided by sustainable development and
 244 combines the innovation of products and production processes with features such as
 245 environmental protection and resource conservation [42, 43] . Summarizing the existing
 246 studies, we can see that the number of corporate patents is a common indicator for
 247 measuring innovation performance[44]. Referring to previous studies, this study measures
 248 the level of green innovation by the number of green patent applications. The green
 249 innovation activities of enterprises can be divided into substantive innovation and strategic
 250 innovation. Invention patent is a kind of substantive innovation achievement which reflects
 251 the enterprises' pursuit of green innovation "quality". In contrast, the utility model patent is a
 252 kind of less difficult strategic innovation, reflecting the pursuit of "quantity" of green
 253 innovation. Specifically, in this study, the total amount of green innovation is the sum of the
 254 number of green invention patent applications and the number of green utility model patent
 255 applications. At the same time, the number of green invention patent applications can be
 256 used to measure the quality of green innovation, and the number of green utility model
 257 patent applications can be used to measure the quantity of green innovation. In order to
 258 eliminate the right-skewed distribution problem of green patent application data, this study
 259 takes the natural logarithm of the number of green patent applications after adding 1 to get
 260 the green innovation indicators LnTotal, LnInva and LnUma.

261 **Table1 The description of variables**

	Variable Abbreviation	Meaning of Variables
Independent variable	Concentration	Degree of population concentration in urban agglomeration
Dependent variable	LnTotal	Total green innovation: $\ln(1+ \text{the number of green patent applications})$
	LnInva	Green innovation quality: $\ln(1+ \text{the number of green invention patent applications})$
	LnUma	Number of green innovations: $\ln(1+ \text{the number of green utility model patent applications})$
Enterprise-level control variables	Cashflow	Company operating cash flow/total assets
	Lev	Asset liability ratio
	ROA	Net profit/total assets

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	Variable Abbreviation	Meaning of Variables
Enterprise-level control variables	PPE	Cash paid for fixed assets, intangible assets and other long-term assets/total assets
	Top1	Shareholding ratio of the largest shareholder
	Seperation	Seperation of two rights
	Age	Year of establishment
	Soe	state-owned enterprises assigned a value of 1
City-level control variables	Fiscal_e	Local fiscal general public budget expenditure/regional GDP
	Loan	Loan balance of depository financial institutions/regional GDP
	Road	road person ratio
	Fdi	Actual utilized foreign capital × average exchange rate / regional GDP
	Scale	Population size of central cities

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267 4. RESULTS AND DISCUSSION

268 4.1.Data and descriptive statistics

269 Table 2 shows the results of descriptive statistics for the sample, including the number of
270 observations, minimum, mean, standard deviation, and maximum. From the result of the
271 descriptive statistics we can see that the standard deviation of Chinese population
272 concentration in urban agglomeration is 0.64 and the mean value is 3.29, which indicates
273 that there are some differences in the degree of population concentration in different urban
274 agglomeration. For the control variables, the mean value of gearing ratio is 41.00%, the
275 mean value of return on assets is about 4% and the percentage of state-owned enterprises
276 is 38%, which are generally consistent with the descriptive statistics in the existing literature.
277

278 **Table 2 Summary statistics**

variable	observations	mean	min	max	sd
LnTotal	19953.00	0.41	0.00	3.85	0.78
LnInva	19953.00	0.27	0.00	3.53	0.61
LnUma	19953.00	0.25	0.00	3.09	0.57
Concentration	19953.00	3.29	1.18	4.33	0.64
Scale	19953.00	6.45	4.71	8.11	0.65

variable	observations	mean	min	max	sd
Cashflow	19953.00	0.05	-2.61	8.85	0.07
Lev	19953.00	0.41	0.05	0.93	0.20
ROA	19953.00	0.04	-0.28	0.20	0.05
PPE	19953.00	0.05	0.00	0.55	0.05
Top1	19953.00	0.35	0.02	0.89	0.15
Seperation	19953.00	4.39	0.00	28.32	7.10
Age	19953.00	16.58	4.00	31.00	5.40
Soe	19953.00	0.38	0.00	1.00	0.49
Fiscal_e	19953.00	0.15	0.07	0.28	0.05
Loan	19953.00	1.52	0.40	3.15	0.63
Road	19953.00	14.95	4.04	33.26	7.30
Sec_ind_ratio	19953.00	0.42	0.16	0.65	0.11
Fdi	19953.00	0.03	0.00	0.12	0.02

279 **4.2. The impact of population concentration in urban agglomeration on**
280 **corporate green innovation**

281 This study investigates the correlation between population concentration and corporate
282 green innovation in urban agglomeration, and the results are shown in Table3. The
283 dependent variable in columns (1) (2) is the total number of green patent applications
284 (LnTotal). The dependent variable in columns (3) (4) is the quality of corporate green
285 invention (LnInva). The dependent variable in columns (5) (6) is the quantity of corporate
286 green innovation (LnUma). Region-level and firm-level control variables are added in
287 columns (2)(4)(6). Additionally, time and industry fixed effects are controlled in all columns.
288 From the result in columns (1)(2), we can see that the regression coefficients of population
289 concentration in urban agglomeration and corporate green innovation are significantly
290 positive at the 1% level, indicating that the increase of population concentration in urban
291 agglomeration can promote corporate green innovation activities and increase the number of
292 corporate green patent applications. Specifically, if the population concentration level of
293 urban agglomeration increases by one unit, the green innovation of enterprises in the
294 domain will increase by 0.06 units. Moreover, from the result in columns (3)-(6), we can see
295 that the regression coefficients are also significantly positive at the 1% level, indicating that
296 the increase of population concentration in urban agglomeration can promote corporate
297 green innovation quality and quantity.

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Table 3 The relationship between Concentration and corporate green innovation

variables	(1) LnTotal	(2) LnTotal	(3) LnInva	(4) LnInva	(5) LnUma	(6) LnUma
Concentration	0.0522*** (0.0161)	0.0623*** (0.0174)	0.0346*** (0.0122)	0.0418*** (0.0134)	0.0419*** (0.0113)	0.0512*** (0.0121)
Scale		0.0137 (0.0200)		0.0145 (0.0156)		0.00417 (0.0140)
Cashflow		0.0754 (0.0834)		0.0801 (0.0649)		-0.00949 (0.0613)

variables	(1) LnTotal	(2) LnTotal	(3) Lnlnva	(4) Lnlnva	(5) LnUma	(6) LnUma
Lev		0.524*** (0.0584)		0.374*** (0.0459)		0.350*** (0.0413)
ROA		1.254*** (0.162)		0.999*** (0.129)		0.682*** (0.114)
PPE		0.326** (0.156)		0.215* (0.123)		0.210* (0.110)
Top1		-0.0321 (0.0733)		-0.0566 (0.0578)		0.0326 (0.0516)
Seperation		0.00387** (0.00154)		0.00282** (0.00119)		0.00224** (0.00110)
Age		-0.00814*** (0.00224)		0.00481*** (0.00169)		0.00555*** (0.00159)
Soe		0.0999*** (0.0279)		0.0968*** (0.0224)		0.0352* (0.0188)
Fiscal_e		0.0804 (0.272)		0.0277 (0.218)		0.177 (0.185)
Loan		-0.0281 (0.0213)		-0.0202 (0.0164)		-0.00830 (0.0149)
Road		0.000323 (0.00187)		0.000842 (0.00145)		-0.000531 (0.00132)
Sec_ind_ratio		-0.180 (0.149)		-0.218* (0.119)		0.0493 (0.102)
Fdi		-0.211 (0.475)		0.159 (0.377)		-0.621* (0.327)
Constant	0.243*** (0.0533)	0.0305 (0.198)	0.159*** (0.0401)	-0.0167 (0.153)	0.109*** (0.0372)	-0.0849 (0.143)
Observations	19,953	19,953	19,953	19,953	19,953	19,953
R-squared	0.146	0.168	0.113	0.135	0.137	0.153

Robust standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

300 4.3. Further Analysis.

301 This study initially explores the positive impact of population concentration in urban
302 agglomeration on corporate green innovation in the domain. However, it is worth noting that
303 there are significant differences between intra-provincial agglomeration and cross-provincial
304 agglomeration. On the one hand, cross-provincial urban agglomeration needs to face the
305 problem from higher-level governments in different provinces, which will increase the
306 difficulty of cross-provincial urban agglomeration development and reduce the positive
307 spillover effect of cross-provincial urban agglomeration development. At the same time, the
308 existence of competitive relationship among provincial governments may also hinder the
309 process of building and developing cross-provincial urban agglomeration. On the contrary,
310 cities within the same province have close geographic location, common customs and
311 convenient conditions for integrated development, which can lay a good development
312 foundation for promoting the construction of urban agglomeration. On the other hand, cross-
313 provincial urban agglomeration can make cities form closely linked socio-economic
314 networks, break the restrictions of administrative divisions, and build a "win-win" and

315 "integrated" development pattern with complementary advantages, resource sharing and
 316 coordination.

317

318 **Table 4 The difference between cross-provincial and intra-provincial concentration**

variables	(1)	(2)	(3)	(4)	(5)	(6)
	LnTotal	LnInva	LnUma	LnTotal	LnInva	LnUma
	Intra-provincial urban agglomeration			Cross-provincial urban agglomeration		
Concentration	0.0619*** (0.0228)	0.0390** (0.0178)	0.0522*** (0.0156)	0.0939*** (0.0319)	0.0746*** (0.0228)	0.0569** (0.0240)
Scale	0.00282 (0.0254)	0.00758 (0.0202)	0.00131 (0.0176)	0.0703** (0.0338)	0.0478* (0.0250)	0.0404 (0.0251)
Cashflow	0.140 (0.128)	0.124 (0.0978)	0.00780 (0.0964)	-0.0175 (0.106)	0.00885 (0.0835)	-0.0465 (0.0768)
Lev	0.511*** (0.0867)	0.338*** (0.0673)	0.377*** (0.0629)	0.539*** (0.0800)	0.418*** (0.0641)	0.313*** (0.0542)
ROA	1.143*** (0.221)	0.881*** (0.172)	0.660*** (0.159)	1.469*** (0.238)	1.188*** (0.193)	0.796*** (0.165)
PPE	0.506** (0.209)	0.346** (0.168)	0.339** (0.146)	0.128 (0.227)	0.0792 (0.177)	0.0694 (0.162)
Top1	-0.158 (0.107)	-0.154* (0.0843)	-0.0632 (0.0751)	0.0194 (0.101)	-0.0132 (0.0800)	0.0670 (0.0715)
Seperation	0.00361* (0.00213)	0.00277* (0.00163)	0.00225 (0.00154)	0.00524** (0.00212)	0.00351** (0.00167)	0.00320** (0.00149)
Age	0.00943*** (0.00312)	0.00587** (0.00231)	0.00605*** (0.00229)	-0.00512 (0.00325)	-0.00202 (0.00256)	0.00451** (0.00222)
Soe	0.101** (0.0406)	0.114*** (0.0328)	0.0160 (0.0269)	0.0985** (0.0382)	0.0761** (0.0304)	0.0586** (0.0263)
Fiscal_e	0.677 (0.418)	0.557 (0.340)	0.506* (0.283)	-0.304 (0.396)	-0.339 (0.311)	-0.0149 (0.275)
Loan	-0.0476 (0.0296)	-0.0377* (0.0227)	-0.0168 (0.0207)	-0.00448 (0.0309)	-0.00339 (0.0240)	0.00176 (0.0216)
Road	0.00118 (0.00274)	0.00201 (0.00219)	-0.00133 (0.00182)	0.00204 (0.00285)	0.000739 (0.00219)	0.00229 (0.00201)
Sec_ind_ratio	-0.288 (0.223)	-0.310* (0.174)	-0.0592 (0.158)	-0.303 (0.224)	-0.307* (0.178)	-0.0474 (0.152)
Fdi	-0.236 (0.733)	-0.0786 (0.582)	-0.427 (0.503)	-0.145 (0.616)	0.296 (0.479)	-0.598 (0.441)
Constant	0.170 (0.270)	0.0866 (0.211)	0.00184 (0.192)	-0.505* (0.285)	-0.370* (0.218)	-0.376* (0.206)
Year FE	yes	yes	yes	yes	yes	yes
Industry FE	yes	yes	yes	yes	yes	yes
Observations	9,996	9,996	9,996	9,954	9,954	9,954
R-squared	0.188	0.154	0.167	0.181	0.146	0.170

Robust standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

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320 In order to better understand the difference between population concentration of intra-
321 provincial urban agglomeration and population concentration of inter-provincial urban
322 agglomeration, this study intends to use sub-sample regression to investigate their
323 relationship with corporate green innovation. Specifically, based on the definition of
324 population concentration in urban agglomeration, if the sample cities in the urban
325 agglomeration all belong to the same province, the study will define these samples as intra-
326 provincial population concentration in urban agglomeration, otherwise they are defined as
327 cross-provincial population concentration in urban agglomeration. The specific regression
328 results are shown in Table 4. The results in columns (1)-(6) show that the coefficients of
329 population concentration in urban agglomeration are significantly positive at the 1% level,
330 indicating that the population concentration in intra-provincial urban agglomeration and
331 cross-provincial urban agglomeration can effectively promote corporate green innovation in
332 the domain. Moreover, from the result of sub-sample regression, this study finds significant
333 differences in the regression coefficients of key explanatory variables. The result shows that
334 compared with population concentration in intra-provincial urban agglomeration, population
335 concentration in cross-provincial urban agglomeration can significantly enhance corporate
336 green innovation level in the domain.

337 **4.4. Heterogeneity analysis**

338 **4.4.1 Regional heterogeneity analysis**

339 Firstly, in the early stage of Chinese reform and opening-up, the eastern region became the
340 pioneer of reform by virtue of its geographical advantage, which made its economic
341 development level and regional innovation vitality better than those of the central and
342 western regions. In order to balance the development gap between regions, China has
343 introduced regional development strategies such as "Western Development" and "Rise of
344 Central China" for the central and western region. In this study, the sample is divided into the
345 eastern region and the central and western region according to the geographical location for
346 group regression. The regression results are shown in columns (1) and (2) of Table 5. In
347 both the eastern and the central and western region, the increase of population
348 concentration in urban agglomeration is beneficial to enterprise green innovation, and the
349 effect is more obvious for the central and western region. Compared with the cities in the
350 eastern region, the cities in the central and western regions are relatively less endowed with
351 innovation resources, and they can promote green innovation development by "grouping"
352 with neighboring cities by forming a synergy.

353 Secondly, in this study, the sample is divided into high marketization group and low
354 marketization group according to the degree of regional marketization. The regression
355 results are shown in columns (3) and (4) of Table 5. In both high marketization and low
356 marketization regions, the increase of population concentration in urban agglomeration is
357 beneficial to enterprise green innovation, and the effect is more obvious for the low
358 marketization regions. Green innovation in low marketization areas is more constrained than
359 in high marketization areas. Coordinated development with neighboring cities is conducive to
360 the improvement of green innovation capacity in low marketization areas, as firms can
361 benefit from the market externalities of the neighboring areas.

362 Thirdly, in this study, cities are classified into large cities and small and medium-sized cities
363 according to the average of city population size during the sample period. The regression
364 results are shown in columns (5) and (6) of Table 5. In both large cities and small and
365 medium-sized cities, the increase of population concentration in urban agglomeration is
366 beneficial to enterprise green innovation, and the effect is more obvious for the small and
367 medium-sized cities. Enterprises located in small and medium-sized cities can enjoy the
368 scale advantage of large markets and accumulate industry-related technical knowledge by

369 interacting with large cities in urban agglomeration, which is beneficial to the development of
 370 corporate green innovation activities.

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372 **Table 5 Regional heterogeneity analysis**

variables	(1)	(2)	(3)	(4)	(5)	(6)
	East	Midwest	High marketization	Low marketization	Big city	Small city
	LnTotal	LnTotal	LnTotal	LnTotal	LnTotal	LnTotal
Concentration	0.0227* (0.0135)	0.0937*** (0.0139)	0.0549*** (0.0121)	0.0895*** (0.0146)	0.0345* (0.0193)	0.0832*** (0.0113)
Scale	0.0225 (0.0175)	0.0512*** (0.0168)	0.0453*** (0.0138)	-0.00169 (0.0183)	0.127*** (0.0364)	0.0206 (0.0173)
control variable	control	control	control	control	control	control
Constant	0.00274 (0.247)	-0.853*** (0.164)	-0.571*** (0.159)	-0.324 (0.228)	1.096*** (0.276)	-0.314* (0.175)
Year FE	yes	yes	yes	yes	yes	yes
Industry FE	yes	yes	yes	yes	yes	yes
Observation	10,186	9,563	10,546	9,203	9,424	10,325
R-squared	0.185	0.175	0.174	0.197	0.177	0.192

Standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

373 **4.4.2 Firm heterogeneity analysis**

374 Firstly, in this study, enterprises are classified into large enterprises and small and medium-
 375 sized enterprises according to the median of enterprise size during the sample period. The
 376 regression results are shown in columns (1) and (2) of Table 6. For both large enterprises
 377 and small and medium-sized enterprises, the increase of population concentration in urban
 378 agglomeration is beneficial to corporate green innovation, and the effect is more obvious for
 379 large enterprises. Enterprises with larger production scale have more significant scale effect
 380 and the ability to layout their businesses in neighboring cities, which allows them to better
 381 take advantage of the growth potential of neighboring markets. In columns (3) and (4) of
 382 Table 6, enterprises are classified into state-owned enterprises and non-state-owned
 383 enterprises according to the nature of the enterprises. The result shows that the effect of
 384 population concentration in urban agglomeration on corporate green innovation is more
 385 significant in the state-owned enterprises compared to non-state-owned enterprises.

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387 **Table 6 Firm heterogeneity analysis**

variable	(1)	(2)	(3)	(4)
	Large enterprises	Small and medium-sized enterprises	State-owned enterprise	Non state-owned enterprises
	LnTotal	LnTotal	LnTotal	LnTotal
Concentration	0.0821*** (0.0143)	0.0312*** (0.0115)	0.0718*** (0.0143)	0.0617*** (0.0122)
Scale	0.0542*** (0.0160)	-0.0389*** (0.0125)	0.0638*** (0.0169)	-0.0125 (0.0130)

variable	(1)	(2)	(3)	(4)
	Large enterprises	Small and medium-sized enterprises	State-owned enterprise	Non state-owned enterprises
	LnTotal	LnTotal	LnTotal	LnTotal
Control variable	control	control	control	control
Year FE	yes	yes	yes	yes
Industry FE	yes	yes	yes	yes
Constant	-0.788*** (0.182)	0.470*** (0.171)	-0.612*** (0.184)	-0.137 (0.177)
Observations	9,768	9,981	7,597	12,152
R-squared	0.231	0.156	0.218	0.176

Standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

388 4.5. Robustness tests

389 Based on the baseline regressions, a series of robustness tests are conducted in this study.

390 4.5.1 Substitution of explanatory variables

391 In terms of the independent variables, I replace the population size data in the process of
392 calculating independent variables. On the one hand, municipal district is an important
393 subdivision of the administrative system in China. It is the area with higher economic activity,
394 which can better reflect the economic development potential of a city and the positive
395 externalities of external radiation. Therefore, the population size of the city is replaced by the
396 population size of the municipal district, and the size of the city itself is adjusted accordingly.
397 The population size of municipal districts is obtained from the China Urban Statistical
398 Yearbook. The regression result in columns (1)-(3) of Table 7 finds that after changing the
399 population size measurement method, the degree of population concentration in urban
400 agglomeration still has a significant positive impact on green innovation of enterprises. On
401 the other hand, given the existence of China's household registration system, many people'
402 household registration and permanent residence locations are different. For example, the
403 urbanization rate of China's household population in 2020 is 45.40%, while the urbanization
404 rate of the resident population is 63.89%. Compared with the registered residence
405 population, the permanent population can better reflect the urban economic vitality.
406 However, due to the fact that China's population census is conducted every 10 years, this
407 study cannot obtain accurate real-time data. Therefore, this study selects the permanent
408 population disclosed in the statistical yearbooks or statistical bulletins of each province as a
409 substitute. The regression result in columns (4)-(6) of Table 7 finds that after changing the
410 population size measurement method, the degree of population concentration in urban
411 agglomeration still has a significant positive impact on green innovation of enterprises.

412 In addition, considering the differences in the sample of municipalities, I exclude the sample
413 firms within the municipalities for robustness testing. The regression result in columns (1)-(3)
414 of Table 8 shows that the regression coefficients of the key explanatory variables are always
415 significantly positive at the 1% level, which again indicates that the above baseline
416 regression results are robust.

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422 **Table 7 Substitution of explanatory variables**

variable	(1) LnTotal	(2) LnInva	(3) LnUma	(4) LnTotal	(5) LnInva	(6) LnUma
Concentration1	0.0433*** (0.0134)	0.0323*** (0.0103)	0.0336*** (0.00956)			
Scale1	0.0207 (0.0163)	0.0144 (0.0128)	0.0132 (0.0114)			
Concentration2				0.0472*** (0.0157)	0.0325*** (0.0123)	0.0387*** (0.0108)
Scale2				0.0418** (0.0210)	0.0316* (0.0164)	0.0267* (0.0145)
Control variable	control	control	control	control	control	control
Constant	0.0719 (0.166)	0.0337 (0.130)	-0.0665 (0.119)	-0.171 (0.200)	-0.146 (0.154)	-0.240* (0.145)
Year FE	yes	yes	yes	yes	yes	yes
Industry FE	yes	yes	yes	yes	yes	yes
Observations	19,953	19,953	19,953	19,953	19,953	19,953
R-squared	0.168	0.135	0.153	0.169	0.135	0.154

Robust standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

423 **4.5.2. Using the dependent variable from the previous period**

424 The baseline regression results in this paper may face endogeneity problem. Although the
 425 independent variable is a city-level variable and the dependent variable is a firm-level
 426 variable, there is still the possibility of reverse causality. For example, as the green
 427 innovation capability of local enterprises grows, it may attract population to the local area
 428 and its surrounding regions, leading to the increase of p population concentration in urban
 429 agglomeration. In addition, the problem faced by baseline regression is that enterprises with
 430 strong green innovation capabilities may prefer to concentrate in areas with high population
 431 concentration. In order to alleviate potential endogeneity issue, considering the potential time
 432 lag of corporate green innovation, this study selects the dependent variable from the
 433 previous period for regression, which results in a temporal mismatch between the
 434 independent variable and the dependent variable. The regression result in columns (4)-(6) of
 435 Table 8 shows that the regression coefficients of the key explanatory variables are still
 436 significantly positive at the 1% level, which indicates that the above baseline regression
 437 results are robust.

438 **Table 8 Using the dependent variable from the previous period**

variable	(1) LnTotal	(2) LnInva	(3) LnUma	(4) F.LnTotal	(5) F.LnInva	(6) F.LnUma
Concentration	0.0698*** (0.0183)	0.0471*** (0.0141)	0.0567*** (0.0127)	0.0701*** (0.0189)	0.0477*** (0.0146)	0.0555*** (0.0132)
Control variable	yes	yes	yes	yes	yes	yes
Constant	0.0109 (0.231)	-0.0452 (0.170)	-0.0437 (0.171)	-0.0367 (0.212)	-0.0659 (0.163)	-0.133 (0.153)
Year FE	yes	yes	yes	yes	yes	yes

variable	(1) LnTotal	(2) Lnlnva	(3) LnUma	(4) F.LnTotal	(5) F.Lnlnva	(6) F.LnUma
Industry FE	yes	yes	yes	yes	yes	yes
Observations	15,717	15,717	15,717	15,930	15,930	15,930
R-squared	0.171	0.139	0.153	0.176	0.142	0.160

Robust standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

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5. DISCUSSION AND CONCLUSIONS

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In this study, I investigate the relationship between population concentration in urban agglomeration and corporate green innovation. My analysis is based on a wide sample of A-share listed companies in china from 2009 to 2019. My empirical results show that (i) population concentration in urban agglomeration has a significant positive impact on corporate green innovation. (ii)The heterogeneous characteristics of a company and area will affect the corporate green innovation capacity. The contribution of population concentration in urban agglomeration to the corporate green innovation is more significant among the central and western regions, the low marketization regions, the small and medium-sized cities, the large enterprises and the state-owned enterprises.

However, there are some limitations in my study which can provide opportunity to further study in the future. The first is the limitation of data source. Compared with listed companies, small companies disclose little information, so the data in this study mainly comes from listed companies, which cannot fully reflect the situation of small companies. Therefore, the results of this study may not be applicable to small enterprises. Green innovation is characterized by the high demand for capital and the high possibility of failure, while small enterprises are characterized by the weak risk tolerance capacity, which will lead to uncertainty about the direction and extent of the impact of population concentration in urban agglomeration on green innovation of small enterprises. Secondly, although my study examines the relationship between population concentration in urban agglomeration and corporate green innovation, I do not further investigate the mechanism underlying this relationship. In future research, more detailed empirical models should be designed to analyze the relevant mechanisms and consequences of higher green innovation capacity.

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COMPETING INTERESTS

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Authors have declared that no competing interests exist.

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