

Learning from the Best: How Regional Knowledge Stimulates Circular Economy Transition at Company Level

Rahel Meili Tobias Stucki Ingrid Kissling-Näf

Bern University of Applied Sciences, Economic Department, Brueckenstrasse 73, 3005 Bern, Switzerland

E-mail: rahel.meili@bfh.ch [<https://orcid.org/0000-0002-1185-278>]; tobias.stucki@bfh.ch; ingrid.kissling@bfh.ch

Abstract: This paper investigates whether, and what kind of, regional knowledge has a stimulating effect on circular economy innovation by companies. We thus add to the literature on regional knowledge spillovers, which has rarely focused explicitly on the circular economy. For the empirical study, we create econometric regressions based on a representative dataset with extensive information on the circular economy activities of about 1,400 Swiss firms. The results confirm that regional knowledge is important for the implementation of circular economy innovations. However, geographical distance and the quality of the knowledge must be taken into account, i.e. companies primarily learn from the best.

Keywords: circular economy; sustainability transition; regional knowledge spillover; anchor tenant; company level data; quantitative analysis.

JEL classification: O30, O33, Q55

Kommentiert [KR1]: Please provide a street address for the affiliation

Introduction

If we want to tackle our environmental problems successfully, there is no way around the circular economy (CE). Around 50% of CO₂ is generated by the materials economy (OECD, 2019). It is therefore very important to make materials management more efficient in the future. This is the goal of the CE: to minimize the amount of waste by using existing resources as long and as efficiently as possible (Esposito et al., 2017). To achieve this goal, a CE makes innovative adjustments to products, services and processes along the entire value chain.

However, we are still in the early stages of the transition to a CE. Every year, 100 billion tons of materials are consumed worldwide; in 2020, only 8.6% of materials were returned to the economy after use (Circle Economy, 2023). The picture is no different at company level. Based on representative company data for Switzerland, Stucki & Woerter (2021) find that only 10% of companies invest significantly in CE activities.

On a positive note, however, there are certainly companies that are already well advanced in CE and show that it can be successfully implemented. These include global companies such as Patagonia, Hilti, Caterpillar and Philips. But there are also such pioneers at regional level. For Switzerland, Stucki et al. (2023) find that there are regional clusters of companies throughout Switzerland that are particularly active in the integration of CE activities.

To understand the prominent role that certain regions play in the transition to a CE, it is essential to look at the specific regional circumstances that support this transformation process. Tapia et al. (2021) emphasize the need for "the identification of spatial regimes and territorial dynamics linked to circular economic transformations". As the transformation to the CE is linked to innovation (Horbach & Rammer, 2020; Scarpellini et al., 2020), we draw on theories of innovation-based regional development to provide explanations. Therefore, we link directly to the literature on regional knowledge spillovers, which argues that firms do not innovate in isolation, but draw on pre-existing external knowledge to create new knowledge.

Kommentiert [KR2]: Journal style does not accommodate the Oxford comma, I have removed them throughout.

Kommentiert [KR3]: Please check this edit for retained sense

49 While this literature is relatively extensive for general innovation (for example: innovative milieux
50 (Camagni, 1991), learning regions (Florida, 1995), regional innovation systems (Autio, 1998), knowledge
51 spillovers (Bottazzi & Peri, 2003), clusters (Porter, 2003), proximities (Boschma, 2005), local buzz (Bathelt
52 et al., 2004), this topic has not yet been empirically investigated for the CE. It is assumed that CE
53 innovations are more complex, as the transition from a linear to a circular business model requires
54 adjustments to existing supply chains, the creation of new networks, the consideration of new
55 technologies and organizational adjustments (e.g., Bocken et al., 2016; De Schoenmakere et al., 2019).

56 Building on the findings from the literature on knowledge spillovers, we also recognize the critical
57 importance of knowledge spillovers from leaders to followers (Eeckhout & Jovanovic, 2002). In addition,
58 the influence of geographical proximity to other companies is recognized as a catalyst for knowledge
59 spillovers (for example Malerba et al., 2013). These earlier findings underline the interconnectedness of
60 spatial dynamics, knowledge transfer and regional clustering in the promotion of CE innovations in
61 specific geographical contexts.

62 In this paper, we want to learn more about the role of leading companies in the transition to a CE and
63 specifically focus on the role of regional knowledge spillovers and how geographical proximity influences
64 the exchange of CE knowledge. This paper therefore aims to make a theoretical contribution to the
65 exploration of how geographical distance influences knowledge spillovers and innovation in the CE
66 context. It also aims to explore the importance of regional CE leaders in facilitating knowledge spillovers
67 at the regional level.

68 For the empirical analysis of these questions, we use one of the world's first quantitative datasets that
69 comprehensively maps the CE at the firm level. The dataset is representative of the Swiss corporate
70 structure and allows for a quantitative analysis of CE activities at the firm level. The data is based on the
71 **KOF (Swiss Economic Institute) Innovation Survey**, which represents a sample of around 8,000 companies.
72 A specific concept was developed for the identification of CE activities (see Stucki et al., 2023), which
73 allows such activities to be measured in a very differentiated manner.

Kommentiert [KR4]: Please define all acronyms at first use

74 **Conceptual background**

75 This section outlines the theoretical background for the role of regional knowledge spillovers
76 in the context of CE innovation. Our exploration will focus on the dual facets of general innovation
77 knowledge and specialized CE knowledge. Furthermore, we will scrutinize the influence of geographical
78 distance on knowledge spillovers and delve into the pivotal role that leading companies play in
79 fostering CE innovation within a given region.

80 CE is a broad concept that explicitly considers activities along the entire production cycle and
81 focuses on the efficient use of existing resources and the closing of resource flows. There is a consensus
82 in the literature that the implementation of CE is quite complex and requires partnerships, alliances,
83 and knowledge linkages (e.g., Bocken et al., 2016; De Schoenmakere et al., 2019). Nevertheless, CE has
84 not yet received much attention in the innovation literature. However, there is a strand that deals with
85 how knowledge affects “green innovations”, i.e. innovations that are associated with a positive
86 environmental effect. While the literature on green innovation focuses heavily on product innovation
87 and the production phase, CE is a broader concept that considers the entire supply chain from
88 procurement to after-use activities (de Jesus & Mendonça, 2018; Geissdoerfer et al., 2018). Due to the
89 complexity of CE innovation, the theories and concepts in the general and green innovation literature
90 are not 1:1 transferable to CE and need to be further specified (Jakobsen et al., 2021). Nevertheless, to
91 formulate hypotheses regarding regional knowledge spillovers in the CE, we primarily turn to the body
92 of literature centred on green innovations. This strand of literature is more closely related than the
93 general innovation literature, as green innovation and CE innovation share the common goal of
94 promoting environmental sustainability, are considered more complex and require more knowledge
95 acquisition than traditional technologies (Barbieri et al., 2020).

96 **Regional knowledge spillovers in the CE**

97 The concepts of innovative milieu, learning regions, clusters, local buzz and proximities share
98 the idea that geographical proximity facilitates and enables collaboration, interaction and the sharing

99 of knowledge and information (Bathelt et al., 2004; Boschma, 2005; Moolaert & Sekia, 2003). In
100 addition, geographical proximity to other firms enables the maintenance of informal relationships with
101 other firms in the region, which is considered one of the main reasons for the spatial concentration of
102 innovation activities (Storper, 1997). Based on these literatures, it can be assumed that regionally
103 available knowledge promotes the implementation of innovations in a company. This is also true for
104 green innovations; for example, Aghion et al. (2016) and Stucki & Woerter (2017) find evidence for
105 regional knowledge spillovers in green innovations. Moreover, the findings of Stucki & Woerter (2022)
106 suggest that external knowledge has a stronger influence on green innovation than non-green
107 innovation. Kok et al. (2021) find that companies that use more external knowledge and adopt more
108 open innovation practices are more likely to adopt green process innovations. As for the specific case
109 of CE, there is little empirical evidence of knowledge spillovers. Stucki et al. (2023) find regional
110 concentrations of firms active in CE in Switzerland. However, to our knowledge, the role of knowledge
111 spillovers has not been directly tested for CE. To shed light on the importance of regional knowledge
112 spillovers for the development of CE activities in companies, we formulate the following hypothesis:

113 **Hypothesis H1:** The more CE knowledge there is in a politically defined region,
114 the more CE innovations a firm will implement.

115 **The role of geographical distance for knowledge spillovers in the CE**

116 As mentioned above, the importance of knowledge interactions for innovation has been
117 extensively discussed and empirically tested. One important aspect that has emerged in this debate is
118 the importance of geographical proximity for the transfer of knowledge and experience. However, the
119 term "regional" is not tied to a specific metric distance and may have different meanings in different
120 national contexts. Absolute distance does matter, however, as research shows.

121 Several studies argue that there is a negative correlation between geographical distance and
122 (green-) knowledge spillovers, meaning that the probability and frequency of knowledge spillover
123 decreases with increasing distance between entities, such as companies or inventors (Malerba et al.,

2013; Murata et al., 2014; Roche et al., 2022; Stucki & Woerter, 2017; van der Wouden & Rigby, 2019; van der Wouden & Youn, 2023). The reasons for these results could be the informal and tacit nature of technological knowledge, which spreads more easily over short distances (Jaffe et al., 1993).

However, the geographic scope should also not be too narrow. If the geographic scope is very narrow, there is a risk that hardly any knowledge is available and consequently no regional knowledge spillover is possible (Glückler et al., 2023). This applies in particular to Switzerland, which in itself already has a very small-scale structure.

So far, there is no evidence on how geographical distance affects knowledge spillovers in the case of CE. Drawing from the insights into the impact of geographical distance on knowledge spillovers, we anticipate the manifestation of an inverted U-shaped relationship, with neither too much geographic restriction (municipal level) nor too much extension (canton level) being ideal for generating knowledge spillovers in the CE. Therefore, we formulate the following hypothesis:

Hypothesis H2: CE knowledge at the district level is more important for the implementation of CE innovations in a focal firm than knowledge at the local or more aggregated cantonal level.

Learning from the best

In this section, we dive deeper into the context of regional knowledge spillover and are interested in the question of whether a leading company in the CE is more valuable than the average number of CE activities that companies pursue in a region. We know from the literature that knowledge spillovers from leaders do flow to so called "followers" (Eeckhout & Jovanovic, 2002). In doing so, they bring new ideas and knowledge to the region, influence other regional actors, and remove institutional, technological, and economic barriers. In this context, Bailey et al. (2018) use the term "anchor tenants." These anchor tenants have a high absorptive capacity, invest in R&D activities and are able to generate knowledge externalities in their region (Agrawal & Cockburn, 2003). They also pioneer the implementation of CE and organize relationships among all other actors (Veyssi re et al., 2022). These

149 anchor tenants build bridges with other actors in the region to align them with their ideas for creating
150 new products and services.

151 Such anchor tenants are considered key components within regional ecosystems to promote
152 sustainable regional competitive advantage (Bailey et al., 2018) and are expected to be relevant to CE
153 ecosystems as well (Aarikka-Stenroos et al., 2020). However, to our knowledge, empirical evidence
154 explaining the role of leaders in a region for the specific case of CE is lacking so far. Therefore, we want
155 to test the following hypothesis:

156 **Hypothesis H3:** The CE knowledge of leading companies, so called "anchor
157 tenants", is more important for the implementation of CE innovations in a
158 focal firm than the knowledge of followers.

159 **The role of general innovation knowledge**

160 There is an extensive strand of literature dealing with knowledge spillovers between different
161 industries. These show that knowledge and experience from different industries are relevant for the
162 development of new ideas, products and processes and that so-called inter-industry knowledge
163 spillovers play a crucial role (Malerba et al., 2013; Scherer, 1982; Verspagen, 1997).

164 Similarly, knowledge from traditional technologies is expected to play a critical role in the
165 development of green innovations (Aghion, Dechezleprêtre, Hemous, et al., 2016). This is especially true
166 because many companies still have little green knowledge and therefore need to build heavily on
167 experience in other areas (Ben Arfi et al., 2018; Stucki & Woerter, 2017, 2022)

168 By combining the insights from these two strands of literature and applying them to our topic of
169 regional sources of knowledge in the CE, we assume that not only CE-specific knowledge, but also
170 general innovation knowledge, will play a role in stimulation CE innovation activities. However, the role
171 of general innovation knowledge for the CE has hardly been studied so far. An exception is Stucki et al.
172 (2023), who find that general innovation knowledge within a company is relevant for the implementation
173 of the CE. We now assume that not only *internal* general knowledge, but also knowledge available in the

174 region can stimulate CE activities. Specifically, we therefore formulate the following hypothesis:

175 **Hypothesis H4:** The more general innovation knowledge there is in a
176 politically defined region, the more CE innovations a company will
177 implement.

178 Empirical concept

179 Data

180 For the empirical analysis, we use unique survey data on CE activities of companies collected in
181 Switzerland in 2020. Switzerland is particularly relevant for this study because, as an innovation-intensive
182 country, it has ideal conditions for a successful transition to a CE and, at the same time, is heavily
183 dependent on such a transition due to its scarce resource deposits. It is therefore likely that the findings
184 from this study can also be transferred to other countries with a time lag.

185 The survey is based on the KOF Enterprise Panel, a stratified random sample of 8,000 Swiss firms that
186 is representative of the population of firms (industry, region, and size class). For the companies in the
187 sample, the panel includes direct contacts who are familiar with completing such questionnaires and are
188 well informed about general business activities; typically, these are the Chief Executive Officer, Chief
189 Financial Officer or Research & Development department head. Since more than 99% of all companies
190 in Switzerland are small and medium-sized enterprises with fewer than 250 employees, these individuals
191 usually have a good overview of the activities in the company and are therefore well suited to complete
192 the questionnaire. To further ensure the representativeness of the data, an extensive round of reminders
193 for underrepresented industries, regions and size classes was carried out after the questionnaires were
194 sent out. Distortions due to self-selection or non-response could thus be minimized. The response rate
195 of 29.1% was satisfactory despite the difficult times during the Covid-19 pandemic (for more information
196 on the composition of the sample, see Stucki & Woerter, 2021).

197 The survey is based on a specially developed concept that distinguishes 27 measures typically relevant
198 to the implementation of the CE at the corporate level. The measures relate to the three fundamental

Kommentiert [KR5]: Please define all acronyms at first use

Kommentiert [KR6]: ibid

199 strategies of a CE and include activities to (a) increase resource efficiency by using fewer resources per
200 product (Bocken et al., 2016), (b) slow down resource cycles by extending product lifetimes and (c) close
201 resource loops through recycling and reuse (McDonough & Braungart, 2010; Stahel, 1994, 1997).
202 Specifically, the survey asked in which of these 27 measures companies achieved measurable
203 improvements between 2017 and 2019 (see Appendix A.1 for specific survey questions). The 27 measures
204 deal with specific sustainability improvements in processes (e.g. logistics, production, procurement) and
205 products (e.g. design) along the entire value chain of the companies and therefore include both CE
206 process and product innovations. These data allow us to ideally map the CE innovation of the companies
207 and, due to the representativeness of the sample, also to derive conclusions about the existing
208 knowledge in different regions. At the same time, the dataset contains further company and market
209 information, such as the intensity of competition, the qualification level of the employees or the industry
210 affiliation, which typically influence the CE activities of the companies (for more information on the
211 survey, see Stucki et al., 2023).

212 **Methodology**

213 Implementing CE activities is essentially a question of innovation capability; at its core, it is about
214 making products and processes increasingly circular, along the entire value chain, through innovative
215 adaptations. Therefore, we build our regression model on a well-founded innovation model (see, e.g.,
216 Cohen, 2010; Crepon et al., 1998). Important factors in such a model are firm size, competition intensity,
217 export activities, financial resources or industry affiliation. Furthermore, in this model it is important to
218 control for specific drivers of the CE such as the strategic embedding of sustainability in the company or
219 energy intensity (see Stucki et al., 2023).

220 We then add the regional innovation knowledge variables to this comprehensive innovation model.
221 To properly identify the effect of regional knowledge, it is critical that we also control for internal
222 innovation knowledge, i.e., firms' absorptive capacity. We do this by controlling for formal knowledge
223 and R&D activities in our baseline model. Thus, our basic regression equation is defined as follows:

$$CE\ activity = \alpha Innovation_drivers + \beta CE_drivers + \gamma Internal_knowledge + \delta Regional_knowledge + \varepsilon \quad (1)$$

where ε is an error term and *Regional_knowledge* distinguishes between (a) the *geographical distance* of regional knowledge (hypothesis 2), (b) the *quality* of regional knowledge (hypothesis 3), or (c) *general* or CE-specific regional knowledge (hypothesis 4), depending on the hypothesis being tested. For a detailed description of all covariates, see the summary statistics in Table 1 and the correlation table in Table A.2.

To empirically investigate the interaction between regional knowledge and the implementation of the CE in companies, we estimate a fractional logit model in a first step. The dependent variable in this model is based on a restricted count variable that can take values between 0 and 27, and was transformed into a fractional variable (Papke & Wooldridge, 1996; Wooldridge, 2010). To examine the extent to which the effect of regional knowledge differs in the transition process, we estimated multinomial logit regressions (Cameron & Trivedi, 2005) in which we defined different groups of firms according to the degree of transformation (Stage 0 (baseline): no CE activities (28% of the companies); Stage 1: 1-4 CE activities (38%); Stage 2: 5-9 CE activities (23%); Stage 3: > or equal 10 CE activities (11%)).

With cross-sectional data, it is almost impossible to identify clear causal effects. However, our main variables relate to information on regional knowledge, which is formed independently of the knowledge of the focal firm. Accordingly, we can assume that the regional variables at best influence the activities of the firms, but are hardly influenced by the activities of the firms themselves. Thus, some causality of effects is already implicit in the data. This is less the case for other model variables. For example, a company's financial resources may promote its CE activities, but it may also be the case that its CE activities gave the company more financial flexibility in the first place. Therefore, although the model includes an extensive vector of control variables that significantly reduces the risk of bias from omitted variables, we refrain from making causal statements and interpret our results primarily as partial

249 correlations.

250 **Empirical results**

251 **Main results**

252 In Table 2, we test the general effects of regional CE knowledge on the implementation of CE
253 innovations in firms. Column 1 shows the results of the fractional logit regression. As expected in
254 Hypothesis 1, these results indicate that regional CE knowledge does indeed have the expected positive
255 correlation with CE innovation of companies. Thus, firms located in regions with more CE knowledge are
256 on average also more likely to introduce CE innovations. In columns 2 to 4, we then estimated a
257 multinomial model that distinguishes different stages of CE innovation implementation. The results
258 suggest that the effect of regional knowledge is relatively linear; regional knowledge thus supports not
259 only the entry into CE (stage 1), but also the frontier of CE innovation (stage 3).

260 In Table 2, we examined the effects of regional knowledge at the district level. However, as expected
261 in Hypothesis 2, geographical distance is likely to play an important role in how large the effect of
262 regional knowledge ultimately is. We examine the role of geographical distance in Table 3. In column 1,
263 we summarize knowledge at a very small level by aggregating it at the level of municipalities; there are
264 2,136 municipalities in Switzerland. In column 2, we do so at the district level, as in Table 2, where the
265 Federal Statistical Office distinguishes 148 districts in Switzerland. In column 3, the aggregation is even
266 broader and is presented specifically for the 26 cantons of Switzerland. As expected in hypothesis 2,
267 knowledge at the district level shows the largest effect. At both the municipality and canton levels, the
268 effect (although positive) is not statistically significant. This result is also confirmed in column 4, where
269 we then simultaneously control for knowledge at the different geographic levels. Not surprisingly, the
270 estimation accuracy decreases somewhat here, since no exclusive categories were formed for the
271 regional knowledge variables (i.e., knowledge at the municipal level is also part of the knowledge at the
272 canton level) and thus the correlation of the three variables is relatively high.

273 In Table 4, we test the effects of the quality of knowledge. Specifically, we want to test whether there

274 are mainly knowledge spillovers from leading firms, as expected in hypothesis 3, or whether there are
275 also spillovers from the experience of firms with low implementation knowledge. To this end, we divide
276 regional knowledge (at the district level) into three different variables: in column 1, we test the effect of
277 low implementation knowledge (proportion of companies with 1-5 out of 27 CE activities in the region
278 of the focal company excluding the focal company), in column 2 the effect of medium implementation
279 knowledge (proportion of companies with 6-10 CE activities), and in column 3, as in the previous
280 regressions, the effect of high implementation knowledge (proportion of companies with more than 10
281 CE activities). Finally, in column 4, all three variables are included in the model simultaneously, with the
282 proportion of firms without CE activities serving as the reference category. In line with hypothesis 3, the
283 results indicate that ultimately only frontier knowledge achieves an effect, i.e., companies can only learn
284 from the best. For deep and medium implementation knowledge, the effects are close to zero and clearly
285 not statistically significant.

286 Finally, to test hypothesis 4, we add a variable to our model that reflects regional knowledge about
287 general innovations (see Table 5). Column 1 suggests that general knowledge also has a positive effect
288 on the implementation of CE innovation at firm-level. However, the effect is somewhat smaller compared
289 to CE-specific knowledge, although the difference is not statistically significant (see column 2).

290 As for the effects of the other model variables, there were few surprises. Consistent with previous
291 results (e.g., Stucki et al., 2023), we find that, in particular, the availability of financial resources, non-price
292 competition, R&D activities, firm size and strategic embedding of sustainability in the firm are important
293 characteristics of firms with CE activities.

294 **Robustness**

295 To check the robustness of our results, we perform several tests (see Table A3). In principle, it is
296 possible that CE knowledge is most widespread in regions where customers are also more aware of
297 sustainability issues and, accordingly, companies are more willing to adopt CE innovations. In principle,
298 we would expect this awareness to be stronger in cities. To test for such possible omitted variable bias,

299 we therefore include controls for whether a company is located in core municipalities or agglomeration
300 municipalities, respectively, in columns 1 and 2. While the effect for core municipalities is clearly
301 statistically significant, it has only a small effect on the effect of regional knowledge.

302 In our main regressions, we test the effect of available knowledge in the region of the focal firm. It is
303 possible that these effects reflect not only the influence of knowledge but also certain regional effects
304 (e.g., different political environments). To control for such effects, we add in column 3 fixed effects for
305 the seven major regions of Switzerland which are defined by the Federal Statistical Office. This reduces
306 the effect of regional knowledge somewhat, but it remains statistically significant and positive. The fact
307 that the effect becomes somewhat smaller is not surprising, since we now distinguish primarily regional
308 differences within the major regions, thus strongly limiting the variance in the knowledge variables.

309 Another potential bias may arise if regional knowledge is tied to specific industry structures, which in
310 turn could affect firms' ability to implement CE innovations. In each of the previous models, we controlled
311 only for sector effects, which can only partially address such a bias. In column 4, we therefore add
312 controls for 34 industry classes. However, this has little effect on the size or significance of the effect of
313 regional knowledge.

314 Finally, as our variables for regional knowledge are not measured at company level but at a more
315 aggregated level, the shown robust standard errors may be biased. To deal with this issue, we also made
316 regressions with clustered standard errors which allow for intragroup correlation. The results indicate
317 that our results are robust to such clustering (results are available on request).

318 **Discussion**

319 Our study makes a valuable contribution to the existing literature on regional knowledge spillovers,
320 specifically emphasizing the significance of distance and anchor tenants (e.g., Bailey et al., 2018; van der
321 Wouden & Youn, 2023; Veysière et al., 2022). While prior research, such as that by Malerba et al. (2013),
322 underscores the importance of regional knowledge spillovers for general innovation, our work extends
323 this focus to the realm of CE innovation. Although some studies have explored the impact of regional

324 knowledge on green innovation (Aghion, Dechezleprêtre, Hémous, et al., 2016; Kok et al., 2021; Stucki &
325 Woerter, 2022), the exploration of CE innovation remains relatively scarce. This gap in the literature is
326 particularly crucial because CE innovation is inherently more complicated and multifaceted and has a
327 direct impact on the use of external knowledge. Our findings indicate that both general and CE
328 knowledge within a region significantly influence the implementation of CE innovation and affect both
329 the entry into CE innovation and the technological frontier. However, the nature of regional knowledge
330 proves to be a critical factor.

331 First, *geographical distance* emerges as a pivotal factor, revealing an inverted U-shaped relationship.
332 Spillover is optimal if the geographical distance is neither too small (at municipal level) nor too large (at
333 cantonal level). While this supports previous studies that emphasize the importance of geographical
334 proximity for knowledge spillovers (Malerba et al., 2013; Murata et al., 2014; Roche et al., 2022; van der
335 Wouden & Youn, 2023), our research introduces a novel insight—an inverted U-shaped relationship—
336 which has not been thoroughly discussed in the knowledge spillover literature. This finding goes beyond
337 the typical focus on urban areas and is in line with studies on innovation in rural areas, which discuss the
338 role of scarcely populated areas for innovation (Glückler et al., 2023).

339 Secondly, the *quality* of knowledge proves to be crucial, with spillovers emanating primarily from
340 knowledge originating from leading firms. Interestingly, firms find limited benefits from the experiences
341 of new entrants in the CE. This aspect advances the discourse on the significance of anchor tenants for
342 CE innovation, providing comprehensive quantitative data to confirm their importance (Agrawal &
343 Cockburn, 2003). Importantly, our study breaks new ground by addressing this topic with a quantitative
344 approach for the role of anchor tenants for CE, complementing existing theoretical and qualitative data
345 (Aarikka-Stenroos et al., 2020; Veysière et al., 2022).

346 Thirdly, not only specific CE knowledge, but also *general* innovation knowledge within a region
347 stimulates a company's CE innovation activities. There is therefore not only a knowledge spillover
348 between regions, but also between technologies. This result is in line with the literature on green

349 innovation, where knowledge spillovers from non-green innovations are also observed (Aghion,
350 Dechezleprêtre, Hémous, et al., 2016; Stucki & Woerter, 2017). This result is of great importance, as most
351 companies have hardly any specific CE knowledge to date. If they can now also draw on general
352 innovation knowledge when transforming to a CE, this will make the transition much easier. However, it
353 must also be clearly emphasized at this point that the effect of CE-specific knowledge is significantly
354 greater, so that companies will not be able to avoid building up specific knowledge in this area.

355 **Conclusion**

356 If we are to achieve our environmental goals, we must massively reduce material consumption, which
357 requires a fundamental shift in our economy to a CE. There is no time to lose in this process, so it is
358 important to use existing knowledge and learn from experience. With this paper, we aim to contribute
359 to a better understanding of how existing regional knowledge leads concretely to new CE innovations
360 and what kind of knowledge is particularly important in this process.

361 These findings have immediate implications for policy making. It turns out that knowledge transfer
362 can play a central role in the transition to CE. Accordingly, it is important to promote knowledge sharing
363 among firms by building networks. Since geographical distance plays a role, regional networks are
364 particularly important for this. The main focus here is on experience from the leading firms. Public
365 procurement can be used to specifically promote such flagship examples. However, it is also important
366 that these examples are then widely communicated and that corresponding experiences are widely
367 disseminated. The finding that general knowledge also has an effect shows that not completely new
368 networks have to be built up, but that existing innovation networks can be specifically adjusted to the
369 CE in order to strengthen the knowledge spillover and accelerate the transition.

370 This study is probably the first to comprehensively examine the role of regional knowledge for CE
371 innovation in companies. The results indicate that this is indeed a relevant and interesting area of
372 research. And there is ample room for future studies. Our study is based on cross-sectional data for
373 Switzerland. To test the robustness of the results, it would be desirable to conduct such a study in the

374 future with panel data and also for other countries. This is the only way to identify and understand
375 differences between countries. The focus of our study is on regional knowledge from other companies.
376 However, from the open innovation literature it appears that there are many other relevant sources of
377 knowledge (customers, universities, etc). It would be important to know more precisely how each
378 knowledge source specifically affects the implementation of CE innovation. In this context, further
379 research could also direct the spotlight towards cluster theory (Porter, 1998). The interlinkages,
380 competitive dynamics and geographical proximity between different companies and institutions,
381 combined with the adaptability of infrastructure, support services and institutional backing to the specific
382 needs of companies, could create an environment that favours regional knowledge spillovers for the
383 creation of CE innovations.

384 In addition to the diffusion of knowledge, various regional factors play a pivotal role in driving the
385 advancement of CE innovations. Notably, a study by Meili and Stucki (2023) suggests that regional
386 household income and public procurement exert positive influences. The question arises as to which
387 other regional factors, such as resource scarcity and supply bottlenecks, act as a catalyst in certain
388 regions and encourage companies to actively engage in the CE.

389 Beyond these considerations, a crucial aspect involves exploring the effectiveness of policy
390 instruments in fostering CE innovations and identifying where their greatest impact lies. In this context,
391 delving into research on smart regulation becomes imperative. Such investigations can facilitate the
392 collaborative efforts of diverse public and private stakeholders in crafting a well-balanced mix of policy
393 instruments conducive to CE initiatives (Gunningham & Sinclair, 2017).

394

395

References

- Aarikka-Stenroos, L., Ritala, P., & Llewellyn, D. W. (2020). Circular economy ecosystems: a typology, definitions, and implications. In S. Teerikangas, T. Onkila, K. Koistinen, & M. Mäkelä (Eds.), *Research Handbook of Sustainability Agency* (pp. 260–276). Edward Elgar Publishing Limited. <https://doi.org/https://doi.org/10.4337/9781789906035.00024>
- Aghion, P., Dechezleprêtre, A., Hemous, D., Martin, R., & Van Reenen, J. (2016). Carbon Taxes, Path Dependency and Directed Technical Change: Evidence from the Auto Industry. *Journal of Political Economy*. <https://doi.org/10.1086/684581>
- Aghion, P., Dechezleprêtre, A., Hémons, D., Martin, R., & Van Reenen, J. (2016). Carbon taxes, path dependency, and directed technical change: Evidence from the auto industry. *Journal of Political Economy*, 124(1), 1–51. <https://doi.org/10.1086/684581>
- Agrawal, A., & Cockburn, I. (2003). The anchor tenant hypothesis: exploring the role of large, local, R&D-intensive firms in regional innovation systems. *International Journal of Industrial Organization*, 21(9), 1227–1253. [https://doi.org/10.1016/S0167-7187\(03\)00081-X](https://doi.org/10.1016/S0167-7187(03)00081-X)
- Autio, E. (1998). Evaluation of RTD in regional systems of innovation. *European Planning Studies*, 6(2), 131–140. <https://doi.org/10.1080/09654319808720451>
- Bailey, D., Pitelis, C., & Tomlinson, P. R. (2018). A place-based developmental regional industrial strategy for sustainable capture of co-created value. *Cambridge Journal of Economics*, 42(6), 1521–1542. <https://doi.org/10.1093/CJE/BEY019>
- Barbieri, N., Marzucchi, A., & Rizzo, U. (2020). Knowledge sources and impacts on subsequent inventions: Do green technologies differ from non-green ones? *Research Policy*, 49(2), 103901. <https://doi.org/10.1016/J.RESPOL.2019.103901>
- Bathelt, H., Malmberg, A., & Maskell, P. (2004). Clusters and knowledge: local buzz, global pipelines and the process of knowledge creation. *Progress in Human Geography*, 28(1), 31–56. <https://doi.org/10.1191/0309132504ph469oa>
- Ben Arfi, W., Hikkerova, L., & Sahut, J.-M. (2018). External knowledge sources, green innovation and performance. *Technological Forecasting and Social Change*, 129, 210–220. <https://doi.org/10.1016/j.techfore.2017.09.017>
- Bocken, N. M. P., de Pauw, I., Bakker, C., & van der Grinten, B. (2016). Product design and business model strategies for a circular economy. *Journal of Industrial and Production Engineering*, 33(5), 308–320. <https://doi.org/10.1080/21681015.2016.1172124>
- Boschma, R. (2005). Proximity and Innovation: A Critical Assessment. *Regional Studies*, 39(1), 61–74. <https://doi.org/10.1080/0034340052000320887>
- Bottazzi, L., & Peri, G. (2003). Innovation and spillovers in regions: Evidence from European patent data. *European Economic Review*, 47(4), 687–710. [https://doi.org/10.1016/S0014-2921\(02\)00307-0](https://doi.org/10.1016/S0014-2921(02)00307-0)
- Camagni, R. (1991). Local 'milieu', uncertainty and innovation networks: towards a new dynamic theory of economic space. In R. Camagni (Ed.), *Innovation Networks* (pp. 121–144). Belhaven Press.
- Cameron, A. C., & Trivedi, P. K. (2005). *Microeconometrics: methods and applications*. Cambridge University Press.
- Circle Economy. (2023). *The circularity gap report 2023*.
- Cohen, W. M. (2010). Fifty years of empirical studies of innovative activity and performance. In N. Rosenberg & B. Hall (Eds.), *Handbook of the Economics of Innovation 1* (pp. 129–213). [https://doi.org/10.1016/S0169-7218\(10\)01004-X](https://doi.org/10.1016/S0169-7218(10)01004-X)
- Crepon, B., Duguet, E., & Mairesse, J. (1998). Research, innovation and productivity: An econometric analysis at the firm level. *Economics of Innovation and New Technology*, 7(2), 115–158. <https://doi.org/10.1080/10438599800000031>
- de Jesus, A., & Mendonça, S. (2018). Lost in transition? Drivers and barriers in the eco-innovation road

- to the circular economy. *Ecological Economics*, 145, 75–89.
<https://doi.org/10.1016/J.ECOLECON.2017.08.001>
- De Schoenmakere, M., Hoogeveen, Y., Gillabel, J., Dils, E., & Manshoven, S. (2019). *Paving the way for a circular economy: insights on status and potentials*. European Environment Agency (EEA).
- Eeckhout, J., & Jovanovic, B. (2002). Knowledge Spillovers and Inequality. *American Economic Review*, 92(5), 1290–1307. <https://doi.org/10.1257/000282802762024511>
- Esposito, M., Tse, T., & Soufani, K. (2017). Is the Circular Economy a New Fast-Expanding Market? *Thunderbird International Business Review*, 59(1), 9–14. <https://doi.org/10.1002/TIE.21764>
- Florida, R. (1995). Toward the learning region. *Futures*, 27(5), 527–536. [https://doi.org/10.1016/0016-3287\(95\)00021-N](https://doi.org/10.1016/0016-3287(95)00021-N)
- Geissdoerfer, M., Morioka, S. N., de Carvalho, M. M., & Evans, S. (2018). Business models and supply chains for the circular economy. *Journal of Cleaner Production*, 190, 712–721.
<https://doi.org/10.1016/j.jclepro.2018.04.159>
- Glückler, J., Shearmur, R., & Martinus, K. (2023). Liability or opportunity? Reconceptualizing the periphery and its role in innovation. *Journal of Economic Geography*, 23(1), 231–249.
<https://doi.org/https://doi.org/10.1093/jeg/lbac028>
- Gunningham, N., & Sinclair, D. (2017). Smart regulation. In P. Drahos (Ed.), *Regulatory Theory* (pp. 133–148). ANU Press. <http://www.jstor.org/stable/j.ctt1q1crtm.16>
- Horbach, J., & Rammer, C. (2020). Circular economy innovations, growth and employment at the firm level: Empirical evidence from Germany. *Journal of Industrial Ecology*, 24(3), 615–625.
<https://doi.org/10.1111/JIEC.12977>
- Jaffe, A. B., Trajtenberg, M., & Henderson, R. (1993). Geographic localization of knowledge spillovers as evidenced by patent citations. *The Quarterly Journal of Economics*, 108.
<https://academic.oup.com/qje/article/108/3/577/1881813>
- Jakobsen, S., Lauvas, T., Quatraro, F., Rasmussen, E., & Steinmo, M. (2021). Introduction to the Research Handbook of Innovation for a Circular Economy. In S. Jakobsen, T. Lauvas, F. Quatraro, E. Rasmussen, & M. Steinmo (Eds.), *Research Handbook of Innovation for a Circular Economy* (pp. 2–11). Edward Elgar.
- Kok, R. A. W., Ooms, W., & Ligthart, P. E. M. (2021). Open innovation and the adoption of environmental process innovations: information source and proximity to partner types. In S. Jakobsen, T. Lauvas, F. Quatraro, E. Rasmussen, & M. Steinmo (Eds.), *Research Handbook of Innovation for a Circular Economy*. Edward Elgar.
- Malerba, F., Mancusi, M. L., & Montobbio, F. (2013). Innovation, international R&D spillovers and the sectoral heterogeneity of knowledge flows. *Review of World Economics*, 149(4), 697–722.
<https://doi.org/10.1007/S10290-013-0167-0/TABLES/10>
- McDonough, W., & Braungart, M. (2010). *Cradle to cradle: Remaking the way we make things*. North Point Press.
- Meili, R., & Stucki, T. (2023). Money matters: The role of money as a regional and corporate financial resource for circular economy transition at firm-level. *Research Policy*, 52(10), 104884.
<https://doi.org/10.1016/j.respol.2023.104884>
- Moulaert, F., & Sekia, F. (2003). Territorial innovation models: A critical survey. *Regional Studies*, 37(3), 289–302. <https://doi.org/10.1080/0034340032000065442>
- Murata, Y., Nakajima, R., Okamoto, R., & Tamura, R. (2014). Localized knowledge spillovers and patent citations: A distance-based approach. *Review of Economics and Statistics*, 96(5), 967–985.
https://doi.org/10.1162/REST_a_00422
- OECD. (2019). Global Material Resources Outlook to 2060. In *Global Material Resources Outlook to 2060*. OECD Publishing. <https://doi.org/10.1787/9789264307452-EN>
- Papke, L. E., & Wooldridge, J. M. (1996). Econometric methods for fractional response variables with an application to 401 (k) plan participation rates. *Journal of Applied Econometrics*, 11, 619–632.

- [https://doi.org/10.1002/\(SICI\)1099-1255\(199611\)11:6](https://doi.org/10.1002/(SICI)1099-1255(199611)11:6)
- Porter, M. (1998). Clusters and the new economics of competition. *Boston: Harvard Business Review*, 76(6), 77–90.
- Porter, M. (2003). The Economic Performance of Regions. *Regional Studies*, 37(6–7), 549–578. <https://doi.org/10.1080/0034340032000108688>
- Roche, M. P., Oetttl, A., Catalini, C., & Hall, M. (2022). (Co-)Working in Close Proximity: Knowledge Spillovers and Social Interactions. In *NBER Working Papers 30120*. <https://doi.org/10.3386/W30120>
- Scarpellini, S., Valero-Gil, J., Moneva, J. M., & Andreus, M. (2020). Environmental management capabilities for a “circular eco-innovation.” *Business Strategy and the Environment*, 29(5), 1850–1864. <https://doi.org/10.1002/BSE.2472>
- Scherer, F. M. (1982). Inter-Industry Technology Flows and Productivity Growth. *The Review of Economics and Statistics*, 64(4), 627. <https://doi.org/10.2307/1923947>
- Stahel, W. R. (1994). *The utilization-focused service economy: Resource efficiency and product-life extension*. National Academy Press, Washington, DC.
- Stahel, W. R. (1997). *The functional economy: cultural and organizational change* (Vol. 1). National Academy Press Washington, DC.
- Storper, M. (1997). *The regional world*. The Guilford Press.
- Stucki, T., & Woerter, M. (2017). Green inventions: Is wait-and-see a reasonable option? *Energy Journal*, 38(4), 43–68. <https://doi.org/10.5547/01956574.38.4.tstu>
- Stucki, T., & Woerter, M. (2021). *Statusbericht der Schweizer Kreislaufwirtschaft. Erste repräsentative Studie zur Umsetzung der Kreislaufwirtschaft auf Unternehmensebene. Schlussbericht im Auftrag des Bundesamts für Umwelt und Circular Economy Switzerland*.
- Stucki, T., & Woerter, M. (2022). Operating successfully on a new technological path: The effect of external search. *Sustainability (Switzerland)*, 14(2), 957. <https://doi.org/10.3390/su14020957>
- Stucki, T., Woerter, M., & Loumeau, N. (2023). Clearing the fog: How circular economy transition can be measured at the company level. *Journal of Environmental Management*, 326. <https://doi.org/10.1016/j.jenvman.2022.116749>
- Tapia, C., Bianchi, M., Pallaske, G., & Bassi, A. M. (2021). Towards a territorial definition of a circular economy: exploring the role of territorial factors in closed-loop systems. *European Planning Studies*, 29(8), 1438–1457. <https://doi.org/10.1080/09654313.2020.1867511>
- van der Wouden, F., & Rigby, D. L. (2019). Co-inventor networks and knowledge production in specialized and diversified cities. *Papers in Regional Science*, 98(4), 1833–1853. <https://doi.org/10.1111/PIRS.12432>
- van der Wouden, F., & Youn, H. (2023). The impact of geographical distance on learning through collaboration. *Research Policy*, 52(2), 104698. <https://doi.org/10.1016/J.RESPOL.2022.104698>
- Verspagen, B. (1997). Measuring Intersectoral Technology Spillovers: Estimates from the European and US Patent Office Databases. *Economic Systems Research*, 9(1), 47–65. <https://doi.org/10.1080/09535319700000004>
- Veyssi re, S., Laperche, B., & Blanquart, C. (2022). Territorial development process based on the circular economy: a systematic literature review. *European Planning Studies*, 30(7), 1192–1211. <https://doi.org/10.1080/09654313.2021.1873917>
- Wooldridge, J. M. (2010). *Econometric Analysis of Cross Section and Panel Data, Second Edition*. The MIT Press.

List of legends:

- Table 1: Descriptive information
- Table 2: The effect of regional CE knowledge on CE activities, Fractional logit (column 1) and multinomial logit regressions (columns 2-4; baseline = stage 0)
- Table 3: Differences by "geographical distance" of regional knowledge, Fractional logit regression
- Table 4: Differences by "quality" of regional knowledge, Fractional logit regression
- Table 5: The role of general regional knowledge, Fractional logit regression
- Table A.1 Survey Questions
- Table A.2: Correlation table
- Table A.3: Including additional controls, Fractional logit regressions

Table 1: Descriptive information

| Variable | Description | Mean | Std. Dev. | Min | Max |
|------------------------------------|--|-------|--------------|------|------|
| Dependent variable | | | | | |
| CE overall | Share of the 27 circular economy activities adopted | 0.16 | 0.16 | 0 | 1 |
| Knowledge variables | | | | | |
| CE know high (district level) | Share of companies with more than 10 CE activities (p90) in the region of the focal company (excluding the focal company) | 0.09 | 0.08 | 0 | 1 |
| General know high (district level) | Share of companies with general R&D expenditures of more than CHF 500,000 (p90) in the region of the focal company (excluding the focal company) | 0.10 | 0.09 | 0 | 1 |
| Control variables | | | | | |
| CE invest | Intensity in circular economy investments | 2.72 | 1.37 | 1 | 6 |
| export | Firm is an exporter, yes/no | 0.43 | 0.50 | 0 | 1 |
| foreign owned | Foreign ownership, yes/no | 0.14 | 0.34 | 0 | 1 |
| price comp | Intensity of price competition, 5-level ordinate variable | 3.95 | 1.02 | 1 | 5 |
| non-price comp | Intensity of non-price competition, 5-level ordinate variable | 3.11 | 1.01 | 1 | 5 |
| R&D | R&D activities, yes/no | 0.31 | 0.46 | 0 | 1 |
| academ education | Share of employees with academic tertiary-level education | 13.95 | 18.09 | 0 | 100 |
| higher education | Share of employees with non-academic tertiary-level education | 16.66 | 15.35 | 0 | 100 |
| apprentices | Share of employees with secondary-level education | 4.96 | 6.30 | 0 | 68 |
| vocational | Share of employees in vocational training | 44.60 | 23.75 | 0 | 100 |
| business-model | Anchoring of sustainability in the business model, transformation of 5-level ordinate variable into binary variable 0 (weak: values 1, 2, 3) and 1 (strong: values 4, 5) | 0.15 | 0.36 | 0 | 1 |
| energy int | Energy cost share, in logs | 0.91 | 0.71 | 0 | 4.51 |
| age | Firm age, in logs | 3.83 | 0.84 | 0 | 6.18 |
| family owned | Firm is family owned, yes/no | 0.56 | 0.50 | 0 | 1 |
| company size | Number of employees measured in full-time equivalents, in logs | 3.95 | 1.31 | 1.61 | 9.82 |

Table 2: The effect of regional CE knowledge on CE activities, Fractional logit (column 1) and multinomial logit regressions (columns 2-4; baseline = stage 0)

| | Stage | CE overall | 1 | 2 | 3 |
|-------------------------------|-------|--------------------|--------------------|--------------------|--------------------|
| CE know high (district level) | | 0.98** (0.40) | 0.47 (0.91) | 1.51 (1.06) | 1.79+ (1.22) |
| CE invest | | 0.24*** (0.02) | 0.52*** (0.08) | 0.78*** (0.09) | 0.87*** (0.10) |
| export | | 0.06 (0.07) | 0.03 (0.18) | 0.29+ (0.20) | 0.03 (0.25) |
| foreign owned | | -0.01 (0.09) | 0.18 (0.24) | -0.06 (0.28) | 0.20 (0.32) |
| price comp | | 0.00 (0.04) | 0.11 (0.08) | 0.15* (0.09) | 0.12 (0.12) |
| non-price comp | | 0.09*** (0.03) | 0.05 (0.08) | 0.09 (0.09) | 0.27** (0.11) |
| R&D | | 0.17** (0.08) | 0.37* (0.20) | 0.49** (0.22) | 0.61** (0.26) |
| academ education | | 0.00 (0.00) | 0.01** (0.01) | 0.02** (0.01) | 0.01 (0.01) |
| higher education | | 0.00 (0.00) | -0.00 (0.01) | -0.00 (0.01) | 0.00 (0.01) |
| apprentices | | 0.01* (0.00) | -0.02 (0.01) | 0.01 (0.01) | 0.02 (0.02) |
| vocational | | 0.00 (0.00) | 0.01+ (0.00) | 0.01** (0.00) | 0.01+ (0.01) |
| business-model | | 0.51*** (0.08) | 0.42+ (0.29) | 1.15*** (0.31) | 1.70*** (0.33) |
| energy int | | 0.12** (0.05) | 0.04 (0.11) | 0.18+ (0.13) | 0.28* (0.15) |
| age | | 0.01 (0.04) | 0.02 (0.09) | 0.10 (0.11) | 0.02 (0.14) |
| family owned | | 0.18*** (0.07) | 0.04 (0.16) | 0.19 (0.19) | 0.30 (0.23) |
| company size | | 0.15*** (0.03) | 0.19*** (0.07) | 0.37*** (0.08) | 0.46*** (0.10) |
| constant | | -3.97*** (0.29) | -2.45*** (0.56) | -5.88*** (0.70) | -7.76*** (0.95) |
| Sector FE | | yes | yes | yes | yes |
| N | | 1404 | | 1404 | |
| pseudo R2 | | | | 0.12 | |
| Wald chi2 | | 403.37*** | | 294.32*** | |
| Log Likelihood | | -433.37 | | -1623.78 | |

Notes: + $p < 0.15$, * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. Stage 0: no CE activities; Stage 1: 1-4 CE activities; Stage 2: 5-9 CE activities; Stage 3: >10 CE activities; robust standard errors in parentheses; sector fixed effects: we include fixed effects for construction, modern services, traditional services, and high-tech manufacturing (reference: low-tech manufacturing).

Table 3: Differences by “geographical distance” of regional knowledge, Fractional logit regression

| | (1) | (2) | (3) | (4) |
|--------------------------------|--------------------|--------------------|--------------------|--------------------|
| | CE overall | | | |
| CE know high (municipal level) | 0.14 (0.17) | | | -0.05 (0.22) |
| CE know high (district level) | | 0.98** (0.40) | | 1.09+ (0.70) |
| CE know high (canton level) | | | 0.34 (0.92) | -0.34 (1.12) |
| CE invest | 0.24*** (0.03) | 0.24*** (0.02) | 0.24*** (0.02) | 0.24*** (0.03) |
| export | 0.00 (0.08) | 0.06 (0.07) | 0.05 (0.07) | 0.01 (0.08) |
| foreign owned | -0.00 (0.09) | -0.01 (0.09) | -0.00 (0.08) | -0.02 (0.09) |
| price comp | 0.02 (0.04) | 0.00 (0.04) | 0.00 (0.04) | 0.02 (0.04) |
| non-price comp | 0.07* (0.04) | 0.09*** (0.03) | 0.09*** (0.03) | 0.07* (0.04) |
| R&D | 0.19** (0.08) | 0.17** (0.08) | 0.17** (0.08) | 0.19** (0.08) |
| academ education | 0.00 (0.00) | 0.00 (0.00) | 0.00 (0.00) | 0.00 (0.00) |
| higher education | -0.00 (0.00) | 0.00 (0.00) | 0.00 (0.00) | -0.00 (0.00) |
| apprentices | 0.01+ (0.01) | 0.01* (0.00) | 0.01* (0.00) | 0.01 (0.01) |
| vocational | 0.00 (0.00) | 0.00 (0.00) | 0.00 (0.00) | 0.00 (0.00) |
| business-model | 0.50*** (0.08) | 0.51*** (0.08) | 0.50*** (0.08) | 0.51*** (0.09) |
| energy int | 0.13** (0.05) | 0.12** (0.05) | 0.11** (0.05) | 0.13** (0.05) |
| age | 0.01 (0.05) | 0.01 (0.04) | 0.00 (0.04) | 0.01 (0.05) |
| family owned | 0.19** (0.07) | 0.18*** (0.07) | 0.18*** (0.07) | 0.18** (0.07) |
| company size | 0.17*** (0.03) | 0.15*** (0.03) | 0.16*** (0.03) | 0.17*** (0.03) |
| constant | -3.98*** (0.32) | -3.97*** (0.29) | -3.92*** (0.30) | -4.01*** (0.33) |
| Sector FE | yes | yes | yes | yes |
| N | 1210 | 1404 | 1411 | 1210 |
| Wald chi2 | 336.51*** | 403.37*** | 387.06*** | 350.07*** |
| Log Likelihood | -374.79 | -433.37 | -436.64 | -374.40 |

Notes: + $p < 0.15$, * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$; robust standard errors in parentheses; sector fixed effects: we include fixed effects for construction, modern services, traditional services, and high-tech manufacturing (reference: low-tech manufacturing).

Table 4: Differences by “quality” of regional knowledge, Fractional logit regression

| | (1) | (2) | (3) | (4) |
|---------------------------------|--------------------|--------------------|--------------------|--------------------|
| | CE overall | | | |
| CE know low (district level) | 0.02 (0.24) | | | 0.28 (0.25) |
| CE know medium (district level) | | -0.03 (0.22) | | 0.21 (0.26) |
| CE know high (district level) | | | 0.98** (0.40) | 1.15*** (0.43) |
| CE invest | 0.24*** (0.02) | 0.24*** (0.02) | 0.24*** (0.02) | 0.24*** (0.02) |
| export | 0.05 (0.07) | 0.05 (0.07) | 0.06 (0.07) | 0.06 (0.07) |
| foreign owned | -0.00 (0.09) | -0.00 (0.09) | -0.01 (0.09) | -0.01 (0.09) |
| price comp | 0.01 (0.04) | 0.01 (0.04) | 0.00 (0.04) | 0.00 (0.04) |
| non-price comp | 0.09*** (0.03) | 0.09*** (0.03) | 0.09*** (0.03) | 0.09*** (0.03) |
| R&D | 0.17** (0.08) | 0.17** (0.08) | 0.17** (0.08) | 0.17** (0.08) |
| academ education | 0.00 (0.00) | 0.00 (0.00) | 0.00 (0.00) | 0.00 (0.00) |
| higher education | 0.00 (0.00) | 0.00 (0.00) | 0.00 (0.00) | 0.00 (0.00) |
| apprentices | 0.01* (0.00) | 0.01* (0.00) | 0.01* (0.00) | 0.01* (0.00) |
| vocational | 0.00 (0.00) | 0.00 (0.00) | 0.00 (0.00) | 0.00 (0.00) |
| business-model | 0.51*** (0.08) | 0.51*** (0.08) | 0.51*** (0.08) | 0.50*** (0.08) |
| energy int | 0.12** (0.05) | 0.12** (0.05) | 0.12** (0.05) | 0.12** (0.05) |
| age | 0.00 (0.04) | 0.00 (0.04) | 0.01 (0.04) | 0.01 (0.04) |
| family owned | 0.18*** (0.07) | 0.18*** (0.07) | 0.18*** (0.07) | 0.18*** (0.07) |
| company size | 0.16*** (0.03) | 0.16*** (0.03) | 0.15*** (0.03) | 0.15*** (0.03) |
| constant | -3.90*** (0.32) | -3.89*** (0.29) | -3.97*** (0.29) | -4.13*** (0.32) |
| Sector FE | yes | yes | yes | yes |
| N | 1404 | 1404 | 1404 | 1404 |
| Wald chi2 | 386.82*** | 387.16*** | 403.37*** | 404.75*** |
| Log Likelihood | -433.93 | -433.93 | -433.37 | -433.27 |

Notes: + $p < 0.15$, * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$; robust standard errors in parentheses; sector fixed effects: we include fixed effects for construction, modern services, traditional services, and high-tech manufacturing (reference: low-tech manufacturing).

Table 5: The role of general regional knowledge, Fractional logit regression

| | (1) | (2) |
|------------------------------------|------------|-----------|
| | CE overall | |
| General know high (district level) | 0.54* | 0.49+ |
| | (0.28) | (0.31) |
| CE know high (district level) | | 0.90** |
| | | (0.41) |
| CE invest | 0.24*** | 0.24*** |
| | (0.02) | (0.02) |
| export | 0.04 | 0.05 |
| | (0.07) | (0.07) |
| foreign owned | 0.01 | 0.00 |
| | (0.09) | (0.09) |
| price comp | 0.00 | 0.00 |
| | (0.04) | (0.04) |
| non-price comp | 0.09*** | 0.09*** |
| | (0.03) | (0.03) |
| R&D | 0.18** | 0.19** |
| | (0.08) | (0.08) |
| academ education | 0.00 | 0.00 |
| | (0.00) | (0.00) |
| higher education | 0.00 | 0.00 |
| | (0.00) | (0.00) |
| apprentices | 0.01** | 0.01* |
| | (0.00) | (0.00) |
| vocational | 0.00 | 0.00 |
| | (0.00) | (0.00) |
| business-model | 0.50*** | 0.50*** |
| | (0.08) | (0.08) |
| energy int | 0.12** | 0.12** |
| | (0.05) | (0.05) |
| age | 0.01 | 0.01 |
| | (0.05) | (0.05) |
| family owned | 0.18*** | 0.18*** |
| | (0.07) | (0.07) |
| company size | 0.16*** | 0.16*** |
| | (0.03) | (0.03) |
| constant | -3.99*** | -4.07*** |
| | (0.30) | (0.30) |
| Sector FE | yes | yes |
| N | 1382 | 1380 |
| Wald chi2 | 383.67*** | 396.21*** |
| Log Likelihood | -426.81 | -424.77 |

Notes: + $p < 0.15$, * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$; robust standard errors in parentheses; sector fixed effects: we include fixed effects for construction, modern services, traditional services, and high-tech manufacturing (reference: low-tech manufacturing).

Table A.1: Survey Questions

Notes: The original questionnaire was conducted in German. The questions presented hereafter were translated by the authors. The original questions in German can be obtained upon request.

Circular business activities

1. During the years 2017-2019, did you achieve any measurable changes within your company in the following areas (I-VII)?

I Procurement yes no

If yes, tick the appropriate boxes; multiple choices possible

(a) Concerning production inputs

- Reducing the ecological footprint of new purchases (e.g., for production, transport)
- Increasing use of used production inputs (up-/downcycling)

(b) In production infrastructure (e.g. buildings, machinery)

- Reducing the ecological footprint of new purchases (production, transport)
- Increasing purchase of used infrastructure
- Increasing purchase of infrastructure with a long product life
- Activities to increase the life span of the production infrastructure used internally (repair, maintenance, etc.)
- Resale of unused infrastructure/materials

II Product/ Service design for customers yes no

If yes, tick the appropriate boxes; multiple choices possible

- Extending product life
- Facilitating repair during use
- Facilitating product updates/upgrades
- Facilitating recycling after use
- Reducing environmental pollution during use/by use (energy consumption, water, soil, air or noise pollution)

III Internal production process yes no

If yes, tick the appropriate boxes; multiple choices possible

- Reducing material consumption (including packaging, paper) in the production process
- Increasing use of renewable energy sources in production
- Reducing environmental pollution in the production process (energy consumption, water, soil, air or noise pollution)
- Reusing waste products and residual materials inside or outside the company

IV Internal storage/logistics yes no

If yes, tick the appropriate boxes; multiple choices possible

- Increasing use of virtualization technology to reduce business travel
- Improving the ecological footprint by optimizing route selection (fuel efficiency) or fleet composition
- Optimizing the logistics/warehouse concept to reduce the required storage space (area and duration)

V Marketing/Sales yes no

If yes, tick the appropriate boxes; multiple choices possible

- Expanding rental/leasing opportunities (Products as a Service)
- Expanding sharing platforms
- Reducing the ecological footprint of correspondence/product documentation

VI After-Sales Services yes no

If yes, tick the appropriate boxes; multiple choices possible

- Extending the warranty or improved maintenance and repair services
- Improving access to spare parts/equipment (lubricants, fuels, batteries)
- Increasing range of product updates/upgrades

VII After-Use Services yes no

If yes, tick the appropriate boxes; multiple choices possible

- Refunds on product returns
- Resale/upgrade of returned products

Table A.2: Correlation table

| | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 12 | 13 | 14 | 15 | 16 | 17 |
|---------------------------------------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|------|------|
| 1 CE overall | 1.00 | | | | | | | | | | | | | | | | |
| 2 CE invest | 0.37 | 1.00 | | | | | | | | | | | | | | | |
| 3 export | 0.09 | 0.09 | 1.00 | | | | | | | | | | | | | | |
| 4 foreign owned | 0.02 | 0.04 | 0.19 | 1.00 | | | | | | | | | | | | | |
| 5 price comp | 0.07 | 0.04 | 0.01 | 0.04 | 1.00 | | | | | | | | | | | | |
| 6 non-price comp | 0.10 | 0.06 | 0.11 | 0.01 | 0.26 | 1.00 | | | | | | | | | | | |
| 7 R&D | 0.17 | 0.16 | 0.40 | 0.03 | 0.02 | 0.11 | 1.00 | | | | | | | | | | |
| 8 academ education | -0.05 | -0.04 | 0.18 | 0.11 | -0.12 | 0.05 | 0.15 | 1.00 | | | | | | | | | |
| 9 higher education | -0.03 | 0.00 | 0.05 | 0.10 | -0.04 | 0.03 | 0.04 | 0.13 | 1.00 | | | | | | | | |
| 10 apprentices | 0.04 | -0.02 | -0.16 | -0.12 | 0.15 | -0.02 | -0.07 | -0.14 | -0.09 | 1.00 | | | | | | | |
| 11 vocational | 0.03 | -0.03 | -0.12 | -0.08 | 0.00 | -0.03 | -0.12 | -0.47 | -0.37 | 0.07 | 1.00 | | | | | | |
| 12 business-model | 0.27 | 0.28 | 0.03 | 0.00 | 0.05 | 0.00 | 0.12 | 0.06 | 0.05 | -0.01 | -0.04 | 1.00 | | | | | |
| 13 energy int | 0.10 | 0.15 | -0.05 | -0.07 | -0.05 | -0.04 | 0.00 | -0.15 | -0.10 | -0.04 | 0.02 | 0.00 | 1.00 | | | | |
| 14 age | 0.10 | 0.11 | 0.07 | -0.04 | 0.13 | -0.01 | 0.03 | -0.15 | -0.03 | 0.13 | 0.10 | -0.02 | 0.02 | 1.00 | | | |
| 15 family owned | 0.07 | 0.04 | -0.01 | -0.14 | 0.19 | 0.03 | 0.00 | -0.26 | -0.13 | 0.08 | 0.10 | -0.03 | 0.08 | 0.10 | 1.00 | | |
| 16 company size | 0.24 | 0.16 | 0.18 | 0.14 | 0.09 | 0.05 | 0.17 | -0.02 | -0.03 | 0.00 | 0.02 | 0.10 | -0.03 | 0.28 | -0.12 | 1.00 | |
| 17 CE know high (district level) | 0.05 | -0.04 | -0.03 | 0.04 | 0.02 | 0.02 | -0.01 | 0.04 | 0.04 | 0.02 | 0.01 | 0.00 | -0.04 | -0.03 | -0.01 | 0.03 | 1.00 |
| 18 General know high (district level) | 0.05 | -0.01 | 0.05 | 0.00 | 0.00 | 0.00 | 0.03 | -0.02 | 0.02 | 0.03 | 0.08 | 0.05 | -0.09 | 0.03 | 0.01 | 0.01 | 0.14 |

Table A.3: Including additional controls, Fractional logit regressions

| | (1) | (2) | (3) | (4) |
|-------------------------------|--------------------|--------------------|--------------------|--------------------|
| | CE overall | | | |
| CE know high (district level) | 0.95** (0.40) | 0.95** (0.40) | 0.75* (0.43) | 0.97** (0.40) |
| CE invest | 0.24*** (0.02) | 0.24*** (0.02) | 0.25*** (0.02) | 0.24*** (0.02) |
| export | 0.04 (0.07) | 0.05 (0.07) | 0.07 (0.07) | 0.04 (0.08) |
| foreign owned | -0.02 (0.09) | -0.02 (0.09) | -0.01 (0.08) | 0.00 (0.09) |
| price comp | -0.00 (0.04) | -0.00 (0.04) | 0.00 (0.04) | -0.01 (0.04) |
| non-price comp | 0.09*** (0.03) | 0.09*** (0.03) | 0.09*** (0.03) | 0.07** (0.03) |
| R&D | 0.17** (0.08) | 0.17** (0.08) | 0.16** (0.08) | 0.17** (0.08) |
| academ education | 0.00 (0.00) | 0.00 (0.00) | 0.00 (0.00) | 0.00 (0.00) |
| higher education | 0.00 (0.00) | 0.00 (0.00) | 0.00 (0.00) | 0.00 (0.00) |
| apprentices | 0.01* (0.00) | 0.01* (0.00) | 0.01+ (0.00) | 0.01* (0.01) |
| vocational | 0.00 (0.00) | 0.00 (0.00) | 0.00 (0.00) | 0.00 (0.00) |
| business-model | 0.51*** (0.08) | 0.50*** (0.08) | 0.50*** (0.08) | 0.53*** (0.08) |
| energy int | 0.11** (0.05) | 0.11** (0.05) | 0.13*** (0.05) | 0.14*** (0.05) |
| age | 0.01 (0.04) | 0.01 (0.04) | -0.00 (0.04) | 0.02 (0.05) |
| family owned | 0.17** (0.07) | 0.17** (0.07) | 0.17** (0.07) | 0.15** (0.07) |
| company size | 0.16*** (0.03) | 0.16*** (0.03) | 0.15*** (0.03) | 0.16*** (0.03) |
| core municipality | -0.15** (0.07) | -0.14* (0.08) | | |
| agglomeration | | 0.02 (0.07) | | |
| constant | -3.92*** (0.29) | -3.93*** (0.29) | -4.04*** (0.29) | -4.30*** (0.46) |
| Sector FE | yes | yes | yes | yes |
| Industry FE | no | no | no | yes |
| Region FE | no | no | yes | no |
| N | 1404 | 1404 | 1404 | 1404 |
| Wald chi2 | 411.86*** | 414.51*** | 426.49*** | 528.97*** |
| Log Likelihood | -433.01 | -433.00 | -432.66 | -430.06 |

Notes: + $p < 0.15$, * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. Robust Standard errors in parentheses; Sector fixed effects: we include fixed effects for construction, modern services, traditional services, and high-tech manufacturing (reference: low-tech manufacturing).