

## **Sustainable supply chain management performance in post COVID-19 era in an emerging economy: a big data perspective**

Nisar Qasim Ali, Haider Shahbaz, Ameer Irfan, Hussain Muhammad Sajjad, Gill Sonaina Safi, Awan Usama

This is a Final draft version of a publication  
published by Emerald  
in International Journal of Emerging Markets

**DOI:** 10.1108/IJOEM-12-2021-1807

### **Copyright of the original publication:**

© 2022, Emerald Publishing Limited

### **Please cite the publication as follows:**

Nisar, Q.A., Haider, S., Ameer, I., Hussain, M.S., Gill, S.S., Awan, U. (2022). Sustainable supply chain management performance in post COVID-19 era in an emerging economy: a big data perspective. International Journal of Emerging Markets. DOI: 10.1108/IJOEM-12-2021-1807

**This is a parallel published version of an original publication.  
This version can differ from the original published article.**



## 44 Sustainable Supply Chain Management Performance in post COVID19 era in an 45 Emerging Economy

46

### 47 1 Introduction

48

49 Planning and integrating digital resources in the supply chain decision-making have become  
50 an environmental necessity for today's firms. The adoption of green supply chain management  
51 strategies strengthens a business's ability to sustain the environment and its financial  
52 sustainability (Khan & Qianli, 2017). Notably, sustainability has emerged as a significant issue  
53 across the industries and particularly united nations are giving adequate attention towards  
54 sustainability and developed several goals for it (United Nations, 2022). Therefore, the  
55 alignment of the green supply chain management (GSCM) and operational management  
56 activities is undeniably important, warrant the supply chain's long-run performance. Big data  
57 and predictive analysis (BDPA) provide the opportunity to manage and understand the supply  
58 chain uncertainties (Bag, 2017). In the ongoing COVID-19 pandemic, manufacturers must  
59 become more mature and resilient in the global supply chain (Ivanov, 2020). Like all other  
60 value chain activities, manufacturing firms have to decide how big data analytics could support  
61 firms in the design of learning and innovation performance (Trabucchi & Buganza, 2018).  
62 Given the nature and importance of big data analytics for supply chain performance, previous  
63 research has called to examine under what conditions a firm improves supply chain  
64 performance (Kamble & Gunasekaran, 2020). Despite this, relatively little is known about what  
65 needs to be done to achieve a long term sustainable supply chain in the COVID-19 environment  
66 (Karmaker et al., 2021) because it has posed several challenges to the supply chain (Alam,  
67 Ahmed, Ali, Sarker, & Kabir, 2021) such as it affected the supply chain momentum  
68 (Karupiah, Sankaranarayanan, Ali, & Paul, 2021) and represents an uncertain situation  
69 (Shukor, Newaz, Rahman, & Taha, 2020). Additionally, the supply chain in emerging markets  
70 is facing several challenges (Yeoman & Santos, 2019).

71

72 Although supply chain risk has come up with many practical suggestions of improving  
73 innovation performance and integrating sustainability practices into the whole supply chain life  
74 cycle (Awan, 2019). There is little knowledge about how supply chain innovativeness improves  
75 sustainability in the supply chain (Bag, Wood, Xu, Dhamija, & Kayikci, 2020). In the current  
76 uncertain environment, the Management of a GSCM could be high risk and decrease the firm's  
77 SC sustainability performance and competitiveness (Mathivathanan, Kannan, & Haq, 2018).  
78 Organizations must reconsider supply chain policy development to prepare for future  
79 pandemics, such as COVID19 (Karmaker et al., 2021) since it has disrupted the supply chain  
80 (Shahed, Azeem, Ali, & Muktadir, 2021). Green supply chain management (GSCM) is widely  
81 regarded as a critical technique for organizations seeking to combine economic, social, and  
82 environmental concerns while promoting organizational sustainability (Ilyas, Hu, &  
83 Wiwattanakornwong, 2020). Now risks are more relevant to be understood and handled with  
84 care as the environmental impact because the organizational actions can aid the risks regarding  
85 reputation and financial insufficiency in emerging countries (De Silva et al., 2021). It is worthy  
86 of mentioning the big data analytics (BDA) tools that can serve as a diver for business benefits

87 and offer an organization improvements. Previous studies have examined supply chain  
88 sustainability determinants, generally adopting big data and predictive analytics capability  
89 (Gunasekaran et al., 2017). While supply chain performance has been previously discussed in  
90 the literature (Dangelico, Pujari, & Pontrandolfo, 2017; Mandal, 2019), some other studies  
91 examined the sustainable developing capabilities (Singh & El-Kassar, 2019) and market  
92 performance (Chakphet, Saenpakdee, Pongsiri, & Jernsittiparsert, 2020) and manufacturing  
93 performance (Dubey, Gunasekaran, & Childe, 2019). According to Siddique et al. (2021), the  
94 recent emergence of the data-driven business markets and exclusion of the conventional data  
95 management systems to track have pushed the application of big data analytics in the supply  
96 chains as well. Bag et al. (2020a) find that big data management analytics positively impacts  
97 innovative green product development and consequently on learning and innovation  
98 performance. Organizations are not unknown to big data analytics as they are growing with the  
99 help of vital data extracted from big data analytics (Salehan & Kim, 2016). Recently, Ali et al.  
100 (2020) contended that few studies are there which address the collection of valuable  
101 information from big data for sustainable product development. Additionally, the development  
102 of the environmental perspectives needs to be considered such as the hazardous quality of life  
103 (Tan & Zhan, 2017). And, the development of social aspects denotes the fulfillment of human  
104 needs by addressing the parties of the innovative product development whereas the economic  
105 perspective deals with the requisites for the economic growth of organizations (Ahmad, Iteng,  
106 Saad, & Abd Rahim, 2018). Recently, Ali et al. (2020) contended that advancing the process  
107 of sustainable development of products influences the value and quality of the products along  
108 with enhancing the image of the organization as well. More importantly, competitive advantage  
109 inform of the sustainable product development can be accomplished by applying the big data  
110 analytics. Since, previous few research studies have examined the impact of big data  
111 management on SC risk management and learning and innovation performance, the present  
112 study has considered both of them as a outcome of the big data analytics capabilities.

113  
114 Accordingly (Lin, Tseng, & Pai, 2018), big data analytics is the tool whose application to  
115 supply chain management of an organization can potentially address the business performance.  
116 Several studies have recently highlighted the importance of big data analytics for supply chain  
117 performance (Bag et al., 2020a). Recently, Sezen (2008) highlighted the necessity to focus on  
118 the impact of supply chain integration and information sharing on supply chain performance.  
119 In this study, we advance the research findings of (Bag et al., 2020a) by introducing SC risk  
120 management in the relationship between big data analytics and innovation and learning  
121 performance. The Research study by (Bag et al., 2020a) has investigated the relationship  
122 between employee development and innovation and learning performance, consequently on the  
123 sustainable supply chain. We modify the conceptual framework offered by (Bag et al., 2020a)  
124 in the manufacturing context by introducing SC risk management practices and providing new  
125 insights into SC literature. Additionally, the previous literature identifying the predictors of the  
126 sustainable supply chain is not much up to date. The previous decade on sustainable supply  
127 chain has extensively emphasized the cost reduction strategies and commercial outcomes for  
128 boosting the supply chains (Liao & Fan, 2020). Notably, the recent pandemic has changed the  
129 old assumptions regarding sustainable supply chains (Choi, 2020). It hints about considering  
130 the new approach for the creation of sustainable supply chains. Karmaker et al. (2021) reported

131 that supply chain leaders are emphasizing emerging technologies to have sustainable supply  
132 chains. Therefore, the present study has addressed the big data analytics capabilities as a  
133 possible driver for the sustainable supply chain in the post-Covid-19 era.

134

135 Motivated by the above-mentioned research gaps and the current situation of Covid-19 the  
136 present study attempted to address sustainable supply chain creation through the big data  
137 analytical capabilities. According to the following (Bag et al., 2020a) research framework, this  
138 study especially investigates whether or not SC risk management mediates the relationship  
139 between big data management capabilities and learning and innovation performance. Notably,  
140 literature is scarce regarding BDA capabilities' role in innovative green product development  
141 and SC risk management, which enhances the learning and innovation that serves as the base  
142 for having a sustainable supply chain. Hence, this study expands the literature discussion on  
143 supply chain performance by answering the following questions. (1) How do different types of  
144 extensive data analytics capabilities influence innovative green product development and  
145 supply chain risk management and consequently on the learning and innovation performance?  
146 (2) Do the innovative green product development and supply chain risk and innovation and  
147 learning performance mediates the association between BDA capabilities and supply chain  
148 performance?

149

150 Because of the increasing focus on cost-based outsourcing and growing concerns of the  
151 stakeholders related to the environment, society, and technology, supply chain sustainability is  
152 considered as important in both the developed and emerging economies (Paul, Ali, Hasan, Paul,  
153 & Kabir, 2022) and sustainability has gained much importance due to the strategic advantage  
154 (Raian et al., 2022). Therefore, the present study examines big data analytics in the Pakistani  
155 context, particularly from Pakistan's multinational companies. Notably, it is quite desperate to  
156 digitize the economy and use big data analytics as well. Recently, government institutions have  
157 made various policies that are directly related to information technology, connectivity, e-  
158 commerce, education, and agriculture, etc. In sum, all of the sectors are being emphasized for  
159 digitization. With a population consisting of 64% of youth (below the age of 30 years) and a  
160 median age of 'digitally savvy' for the youth demographic in Pakistan is 22 years, promise the  
161 digitization. It can be argued that due to the absence of the worldwide big data country, Pakistan  
162 has lagged behind almost 5-10 years in big data (Latif et al., 2018). Additionally, sustainability  
163 performance and dynamic capabilities are improved at the business process level, and it also  
164 happens while managing an organization's resources.

165

166 The study has three-fold contributions to the existing literature. First of all, it determines the  
167 factors such as big data analytics capabilities as a potential predictor for sustainable supply  
168 chain performance. Things do not happen in a vacuum, so the study also identifies and  
169 examines the process of how an organization employing big data analytics can have sustainable  
170 can supply chain performance. Secondly, the study is based on the theoretical gaps identified  
171 by the previous studies (Bag et al., 2020a). Their study contended that further studies might  
172 enhance the scope of BDA by considering the three types of BDA capabilities. Second, a recent  
173 study (Dubey et al., 2019) considered the big data analytics capability as a one-dimensional  
174 construct that may have hindered understanding. Additionally, they also recommended using

175 theories other than contingency theory and dynamic capabilities view. Considering the research  
176 mentioned above, the present study examined the three types of big data analytics capabilities.  
177 It also considered the multinational companies from multiple sectors to have better and broader  
178 insights regarding the influence of significant data analytics capability on the sustainable  
179 supply chain performance. Thirdly, the study is a valuable addition to the literature as it offers  
180 Asian cultural insights on big data analytics, and particularly, to the best of our knowledge, it  
181 is among the few studies. Further, to best of the only research in Pakistan which has provided  
182 the insights regarding the big data analytics and its contribution towards the supply chain risk  
183 resilience and its sustainable performance from the multinational companies related to different  
184 sectors.

185

## 186 **2 Literature review**

187

### 188 **2.1 Theoretical Background**

189 The present study is underpinned by the Dynamic Capability View (DCV). According to  
190 dynamic capability, the theory is an extension and reaction against the inability of the resource-  
191 based view for the interpretation to develop and redevelop the resources and capabilities to  
192 entertain the dynamic environment (Bleady, Ali, & Ibrahim, 2018). Similarly, it is regarded as  
193 the extension of the resource-based view that explains the sustainability of the competitive  
194 advantage for an organization. These capabilities offer the individuals a competitive advantage  
195 (Mukhtar, Baloch, & Khattak, 2019). An organization can easily obtain a sustainable  
196 competitive advantage when they get the dynamic capabilities for gaining the functional  
197 competencies (Pervan, Curak, & Pavic Kramaric, 2017).

198 Besides the explanation of the factors that result in gaining the competitive advantage The  
199 theory also serves to explain the underlying factors that may result in sustainable supply chains.  
200 Accordingly, contended that supply chains do have dynamic changes as compared to the casual  
201 markets due to the changing customer's behaviors or the non-governmental institutions (Hall,  
202 2000). These capabilities help an organization to identify the opportunities in the market and  
203 avail them. Dynamic capabilities theory has been previously used for examining the various  
204 indicators of the supply chains. Accordingly, the present study also adopts the dynamic  
205 capability theory to underpin the current research framework.

206 The study has considered big data analytics as a dynamic capability that enables an organization  
207 to identify the opportunities and avail them. Qwabe (2020) reported that big data as an  
208 analytical capability enables managers, statisticians, trend analysts, and other professionals to  
209 systematically analyze the incoming data. Additionally, the capabilities of an organization  
210 facilitate the productive and competitive utilization of the resources whether they are tangible  
211 or intangible. Thus, the present study argues that different big data capabilities (Management,  
212 talent, and technological) enable an organization to identify and gain sustainable opportunities  
213 that result in innovative green product development and supply chain risk management leading  
214 towards innovation learning and performance enabling the sustainable supply chain.

215

### 216 **2.2 BDA Management Capabilities and Innovative Green Product Development**

217

218 The COVID-19 virus has had a significant impact on health, economic, social, and industrial  
219 activities. It has severely interrupted supply chain management and hampered the transfer of  
220 critical commodities (Karuppiyah, Sankaranarayanan, Ali, & Paul, 2021). Due to an extensive  
221 level of environmental uncertainty, the adoption of developed processes by using big data  
222 capabilities such as management, talent, and technological capabilities is vital for the  
223 organization to attain the supply chain's sustainability performance (Janssen, van der Voort, &  
224 Wahyudi, 2017). Moreover, it claimed that there is an extensive requirement of the initiatives  
225 and to implement the BD programs (Braganza, Brooks, Nepelski, Ali, & Moro, 2017).  
226 Moreover, Gunasekaran et al. (2017) explored the resource-based view to elaborate on the  
227 influence of resources on BD integration capabilities. They argued that the supply chain and  
228 overall firm performance are influenced by the IGPD which is further reliant on the mediation  
229 impact of management support. The success of the IGPD is only realized when a corporation  
230 incorporates operational excellence to attain a competitive advantage. Thus, the managerial,  
231 talent, and technological capabilities help the IGPD achieve sustainable achievements and  
232 growth. Teece (2016) defined that the dynamic capabilities as 'the skills of personals to build,  
233 integrate, and reconfigure competencies to answer rapidly changing environments.' Thus, the  
234 changing environment requires the employees to have Management, technological, and talent  
235 capabilities to incorporate climate change in the organization.

236 The dynamic variation in technology forces corporations to implement strategic deliberation  
237 of personal capabilities such as Management, talent, and technological capabilities. The  
238 difficulties in developing employees associated with contemporary corporations can be  
239 increased by the involvement of an extensive volume of big data (Shah, Irani, & Sharif, 2017).  
240 In addition, Tiwari, Wee, and Daryanto (2018) also conducted studies on BDA related to supply  
241 chain over procurement and strategic sourcing through logistics activities and demand planning  
242 for six years. To achieve the benefits of BDA, particularly for planning, specialized skilled  
243 staff is often required. Likewise, a study conducted by Zhong, Newman, Huang, and Lan  
244 (2016) examined that the application of BD in the business process can enhance the firm's  
245 performance. The technology processing will sharpen the processor to complete the numerous  
246 processing modes to fit various processing development (Mishra, Gunasekaran, Papadopoulos,  
247 & Childe, 2018). A lot of opportunities are available for holistic personals development that is  
248 based on the capabilities of BDA (Wang, Gunasekaran, Ngai, & Papadopoulos, 2016). Thus,  
249 BDA capabilities such as technological and talent enhance employee development, improving  
250 the organization's working style and creating innovative employee competencies (Zhao, Liu,  
251 Zhang, & Huang, 2017). Thus, based on these kinds of literature, the existing study formulates  
252 the following hypotheses:

253

254 ***H1: There is a significant relationship between the BDA management capabilities and***  
255 ***innovative green product development.***

256 ***H2: There is a significant relationship between the BDA talent capabilities and innovative***  
257 ***green product development.***

258 ***H3: There is a significant relationship between BDA technological capabilities and***  
259 ***innovative green product development.***

260

### 261 **2.3. BDA Capabilities and Supply Chain Risk Management**

262 One of the reasons behind BDA development is to provide coverage to aspects of Management,  
263 especially in supply chain management. Such as it provides coverage to the supply chain in  
264 multiple phases, i.e., identifying the prospect, supplying and distributing the product, and  
265 managing the activities of suppliers. The failure of proper management of the supply chain  
266 ends with project collapse. BDA results in the creation of substantial opportunities to develop  
267 the supply chain in the context of worldwide clients. Predictive analytics can forecast supply  
268 chain risk, and true forecasting results in inappropriate marketing and operations section  
269 strategies and cost-cutting in the supply chain process. Organizations apply BDA tools to attain  
270 a competitive advantage. Failure of the supply chain objectives can be eradicated (Zhan, Tan,  
271 Li, & Tse, 2018).

272 Talent capabilities require time and capital investment to develop programming skills, projects,  
273 data, and network management, its maintenance and analytics also manage the supply chain  
274 risk (Akter, Wamba, Gunasekaran, Dubey, & Childe, 2016). Active leaders play a vital role in  
275 generations. As the world has become a global village, the more and more changes in internet  
276 investment, technological skills are essential for existing workers' development and supply  
277 chain activities management in the organization. In agreement with a previous study (Marshall,  
278 Mueck, & Shockley, 2015), investigation leaders emphasize collaboration within the context  
279 of a structured approach by focusing on innovation.

280

281 *H4: BDA management capabilities are positively associated with supply chain risk*  
282 *management.*

283 *H5: BDA talent capabilities are positively associated with supply chain risk management.*

284 *H6: BDA technological capabilities are positively associated with supply chain risk*  
285 *management.*

286

### 287 **2.4. Innovative Green Product Development and Innovation and Learning Performance**

288

289 Although sustainable supply chain performance depends on innovation, the question here is its  
290 implementation to upkeep performance? The true Management and talent capabilities make  
291 innovation (a key player in today's competitive world to secure competitive advantage)  
292 possible. By employing firm Resource Base View (RBV) Wamba et al. (2017) proposed the  
293 BDA leverage model by providing evidence in the livelihood of dynamic capabilities (process-  
294 oriented) to augment the capabilities of the organization to bring improvement in organization  
295 performance. For better development of capabilities, an organization will have to bring together  
296 the complexity of different employees' work and supply chain processes in multiple  
297 organizations (Wang et al., 2016). Further, organizations suggested that in the successful  
298 application of BDA, talent, process, and technological capabilities all together play a vital role.  
299 The development of the innovative green product ends at product launching by having started  
300 with an idea. Environmental knowledge sharing is essential among the environment and  
301 product team specialists to develop an innovative green product in small markets. Proper



302 implementation of the IGPD process results in not only improvement of innovation but also  
303 organization employees' learning performance by permitting debate, errors, and failure analysis  
304 to the development of IGPD project by employing the application of BDA all the levels in that  
305 particular firm (Sivarajah, Kamal, Irani, & Weerakkody, 2017). Notably, the BDA capabilities  
306 at both the operational and strategic levels generate new knowledge and business value.  
307 Following are the hypothesis;

308 ***H7: There is a significant relationship between innovative green product development***  
309 ***innovation and learning performance***

310 ***H8: Innovative green product development is a significant mediator between the relationship***  
311 ***of BDA management capabilities and innovation and learning performance***

312 ***H9: Innovative green product development is a significant mediator between the relationship***  
313 ***of BDA talent capabilities and innovation and learning performance***

314 ***H10: Innovative green product development is a significant mediator between the***  
315 ***relationship of BDA technological capabilities and innovation and learning performance***

316

## 317 **2.5 Supply Chain Risk Management and Innovation & Learning Performance**

318

319 We can characterize the development of the supply chain with multiple events. The  
320 augmentation of continuous technological learning is one of the core factors. In progressive  
321 nature organizations, the personnel associated with the supply chain and committed to  
322 continuing their training usually get higher management support. The innovation and learning  
323 process supports effective supply chain processes (Awan, 2019). In contrast, development  
324 activities support the supply chain in optimizing the resources and boosting supply chain-  
325 related capabilities (Shohan et al., 2019).

326 In the development of competitive advantage, innovation permits the firms in supply chain  
327 development. Innovation is the key player in the process of competitive advantage in such a  
328 progressive culture (Awan et al., 2021). Improvisation in the supply chain process ends with  
329 employee satisfaction along with an increase in organization profit. The opportunities for  
330 growth and development result in strengthening supply chain performance (Moktadir, Ali,  
331 Rajesh, & Paul, 2018). The supply chain can also result in modifying the relationship between  
332 BDA capabilities and an organization's learning.

333 ***H11: There is a significant relationship between the supply chain risk management and***  
334 ***innovation and learning performance***

335 ***H12: Supply chain risk management is a significant mediator between the BDA***  
336 ***management capabilities and innovation & learning performance***

337 ***H13: Supply chain risk management is a significant mediator between BDA talent***  
338 ***capabilities and innovation & learning performance***

339 ***H14: Supply chain risk management is a significant mediator between BDA technological***  
340 ***capabilities and innovation & learning performance***

341

## 342 **2.6 Innovation & Learning Performance and Sustainable Supply Chain Performance**

343

344 One of the labor concentrated industries is the Mining industry, having high co-operation  
345 among employees is predictable as our world has become a global village. Rahman, Ali,  
346 Moktadir, and Kusi-Sarpong (2020) argued that insufficient technology and inadequate  
347 knowledge could also adversely affect a firm to improve GSCM. Alternatively, process  
348 improvement can increase learning, and in turn, can improve GSCM (Chowdhury, Ali, Paul,  
349 Mahtab, & Kabir, 2020). Learning continuously is the key to maximum performance. The  
350 Mining business situation is required the dynamic environment having the necessity of learning  
351 from past instances to avoid the same in the future to contribute to SSCP enhancement  
352 (Hammervoll, Jensen, & Beske, 2012). Innovation can reduce supply chain process costs and  
353 increase organizational profitability (Mahesar, Chaudhry, & Tariq, 2017).

354

355 *H15. There is a significant relationship between innovation & learning performance and*  
356 *sustainable supply chain performance*

357 *H16. Innovation & learning performance is a significant mediator between the relationship*  
358 *of innovative green product development and sustainable supply chain performance*

359 *H17. Innovation & learning performance is a significant mediator between the relationship*  
360 *of supply chain risk management and sustainable supply chain performance*

361

## 362 **2.7 Supply Chain Innovativeness**

363

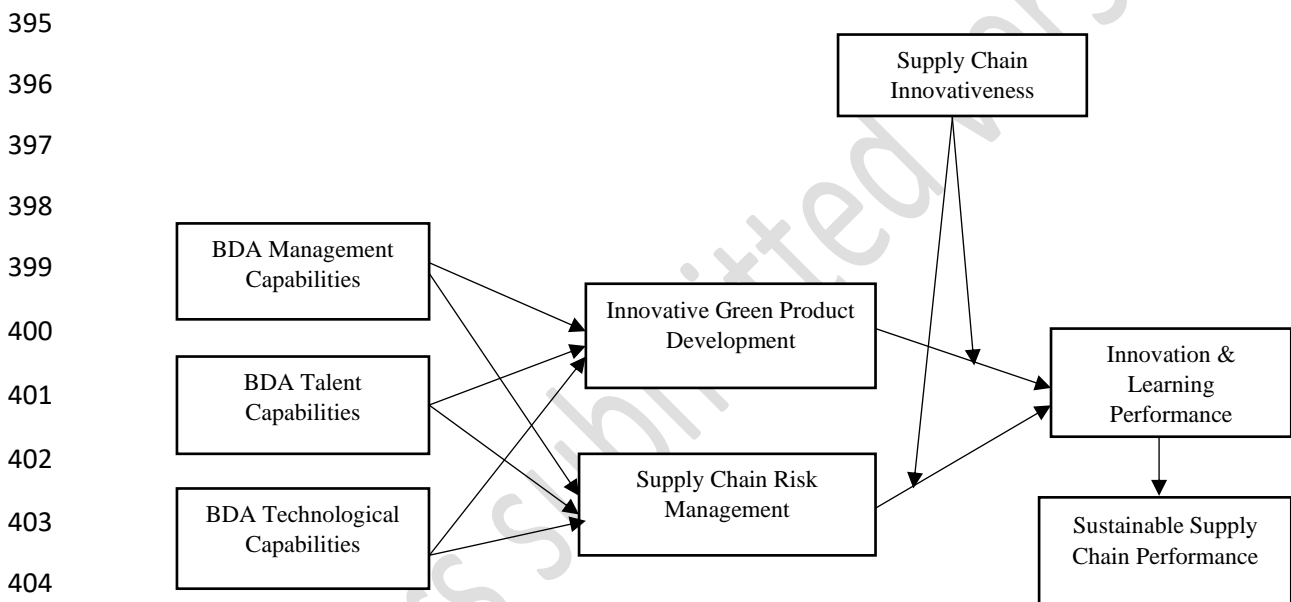
364 Competitive edge is a key player in the process of organization success, having dependence on  
365 innovation. Additionally, innovation throws a direct impact on any organization's supply chain  
366 process (Hult, Hurley, & Knight, 2004). An innovation-based supply chain process of any  
367 organization throws its impact on the new product and the process development while  
368 emphasizing an immediate response to the customer's requirement. In addition, both an  
369 organization's performance and supply chain are at a higher level backed by re-utilization and  
370 assimilation of BDA. Keeping the current technological transformation and intervention in our  
371 routine activities, one can't imagine looking into the future. Sivarajah et al. (2017) having their  
372 investigation on challenges faced by the organization along with BDA and proposed two prime  
373 challenges. 1) Data management process, specifically concerning data acquisition, storage, and  
374 mining, in the requirement of cleansing activities. 2) Challenges faced like Management of  
375 data privacy, its scrutiny and governance, the occurrence of sharing, and finally, Management  
376 and ownership of the data. It is not viable to produce the final product and services by  
377 overseeing innovation base factors in the process of the supply chain (Hult et al., 2004). One  
378 of the recent studies identified the crucial role of supply chain innovativeness to improve  
379 innovation & learning performance. Findings of their study revealed that supply chain  
380 innovativeness significantly moderates the association and strengthens the existing relationship  
381 (Bag, Wood, Xu, Dhamija, & Kayikci, 2020b). Alike, Begum, Ashfaq, Xia, and Awan (2022)  
382 contended that green innovation is an important factor for developing sustainability. So it is

383 asserted that innovative green product development is also necessary for innovation and  
384 learning performance enhancement. The present study proposed that supply chain  
385 innovativeness may moderate the nexus among innovative green product development and  
386 innovation & learning performance based on the above-mentioned arguments. Thus, the  
387 following hypotheses are proposed:

388 *H18. Supply chain innovativeness significantly is a significant moderator between the*  
389 *relationship between innovative green product development and innovation & learning*  
390 *performance.*

391 *H19. Supply chain innovativeness significantly is a significant moderator between the*  
392 *relationship between supply chain risk management and innovation & learning*  
393 *performance.*

394 Figure 1 represents the research framework of the present study.



405  
406  
407  
408 **Figure 1: Research Framework**

### 411 3 Methodology

#### 412 3.1 Population and Sample Size

413 Pakistan is a developing country and desperate for the digitization of the economy. Having said  
414 that, the present study aimed to examine how the big data analytics capabilities contribute  
415 towards sustainable supply chain performance? Moreover, the study also examined innovative  
416 green product development, supply chain risk and innovation, and learning performance as

417 facilitators for creating a sustainable supply chain performance. Big data analytics requires  
418 huge investments and infrastructures; therefore, the present study has considered the  
419 multinational organizations working itself in Pakistan and companies whose subsidiaries are  
420 working in Pakistan. Multinational organizations are selected because they are extensively or  
421 to some extent using big data as compared to the majority of the local (Pakistani) organizations.  
422 According to a State Bank of Pakistan (SBP) report, it provided a list of 899 multinational  
423 companies.

424 The sample size of the study was calculated by using the formula proposed by (Yamane, 1967).  
425 Following is the formula for sample size calculation:

$$426 \quad n = \frac{N}{1 + N(e)^2}$$

427 In this formula n= Sample size; N= Total population; e=Precision level

$$428 \quad n = \frac{899}{1 + 899(.05)^2} = 276$$

429 As per the formula, the minimum sample size for the present study is 276 respondents.  
430 Therefore, the sample size was inflated by 50% to gain the maximum responses from the  
431 respondents. Hence the sample size for the current study is 414 respondents. The sample size  
432 for the current study is also justifiable according to previous studies; for instance, to perform  
433 the structural equation model, a sample size of 200 to 400 respondents is enough (Oke,  
434 Ogunsami, & Ogunlana, 2012). According to Hair, Black, Babin, and Anderson (2010) 100 is  
435 the minimum sample size when there are five or fewer constructs in the model. Hence, there is  
436 no issue with the sample size of the present study.

437

### 438 **3.2 Pre-testing and Measures**

439

440 All of the measures for the present study are adapted from the previous studies. Big data  
441 analytics management capability measured by six items measuring instrument, big data  
442 analytics talent capability measured by nine items measuring instrument (Akter et al., 2016).  
443 Innovative green product development measured by 10 items (Pujari, Wright, & Peattie, 2003),  
444 innovation and learning performance measured by eight items (Akgün, Ince, Imamoglu,  
445 Keskin, & Kocoglu, 2014; Silvestre, 2015), and supply chain risk measured by six items (Singh  
446 & Singh, 2019) and finally sustainable supply chain performance measured by eight items  
447 (Gunasekaran et al., 2017).

448 After finalizing the measuring instrument for variables, it was then presented to two assistant  
449 professors and two practitioners for pre-testing. The pre-testing affirmed the content validity  
450 of the measuring instrument. Data were collected by using the questionnaires. An email was  
451 sent to the senior managers of randomly selected organizations with the purpose of the study  
452 and requested data collection participation. After gaining permission for data collection  
453 questionnaires were distributed among the individuals. Later on, questionnaires were

454 recollected after 3 weeks. Overall, researchers received a total of 200 usable responses from  
455 the respondents and yielded a response rate of 48%. It is considered a good response rate as  
456 previous studies have reported a low response rate. For instance, Dubey et al. (2019) in their  
457 study, reported a response rate of 23.22%. Similarly, Shamim, Zeng, Shariq, and Khan (2019)  
458 also reported a response rate of 27% in their study and also applied the PLS-SEM. PLS-SEM  
459 was employed for data analysis using Smart PLS 3 as it is deemed a popular and advanced  
460 estimation technique in the hospitality and tourism domain (Ali, Rasoolimanesh, Sarstedt,  
461 Ringle, & Ryu, 2018). Moreover, the PLS algorithm followed by bootstrapping technique was  
462 used to assess the measurement and structural model.

## 463 **4 Results**

### 464 **4.1 Demographics**

465 This section deals with the demographic characteristics of respondents. Respondents'  
466 demographic profiles are presented in Table 1. A total of 374 respondents participated in this  
467 survey. Table 1 shows that out of 374 participants, 55.3% (207) were male and 44.7% (167)  
468 were female. The findings reported that the majority of the respondents were aged from 26-45  
469 years. The survey accounted for 67.1% (251) of individuals being in that age group. While  
470 24.9% (93) were of ages up to 25 years old, the remaining 8% (30) belonged to the age group  
471 of 46-55 years. In terms of qualification, 74.6% (279) of participants had bachelor's degree,  
472 followed by 21.4% (80) with master's degree, 2.4% (9) fell in the others' category (Diploma or  
473 Professional Qualification), and the remaining 1.6% (6) had obtained their Ph.D. Meanwhile,  
474 56.7% (212) of respondents had permanent job positions, while 40.1% (150) were on the  
475 contractual nature of employment and the remaining 3.2% (12) of leaders were Internees in  
476 their respective organizations. The results relating to the length of service found that 21.1%  
477 (79) of individuals had up to one-year job experience, 47.9% (179) of respondents had 2-5  
478 years of experience, while 19.3% (72) had 5-10 years of experience, and remaining 11.8% (44)  
479 had 10+ years of job experience in their respective organization.

480 Table 1

481 *Demographics*

482 Insert Table 1 here

483

### 484 **4.2 Measurement Model Assessment**

485 The measurement model was evaluated and convergent validity was assessed by using  
486 loadings, average variance extract, and competitive reliability. Refer to Table 2, except for a  
487 few values; factor loadings exceeded the recommended value of 0.60 for both countries.  
488 Similarly, all values of composite reliability (CR) also exceeded the recommended value of  
489 0.70. All the values of average variance extract (AVE) for all under-study constructs exceeded  
490 the recommended values of 0.50 (Hair Jr, Hult, Ringle, & Sarstedt, 2021). Items were deleted  
491 with the lowest factor loadings (<0.50).

492

493 Table 2

494 *Convergent Validity*

495 Insert Table 2 here

496 **4.3 Discriminant Validity**

497 Henseler, Ringle, and Sarstedt (2015) proposed a new and advanced criterion (HTMT ratio) to  
 498 assess the discriminant validity, and they agreed that the Fornell-Larcker criterion was one of  
 499 the effective methods to evaluate discriminant validity, but this approach had not detected the  
 500 lack of discriminant validity in various research situations. Therefore, the HTMT ratio was  
 501 used to assess the discriminant validity of constructs. As shown in Table 3, all the HTMT ratio  
 502 values were given in Malaysian and Pakistani contexts. All the values are less than 0.90 as  
 503 recommended by Gold, Malhotra, and Segars (2001); hence discriminant validity had also  
 504 been established for all constructs.

505 Table 3  
 506 *Discriminant Validity (HTMT Ratio)*

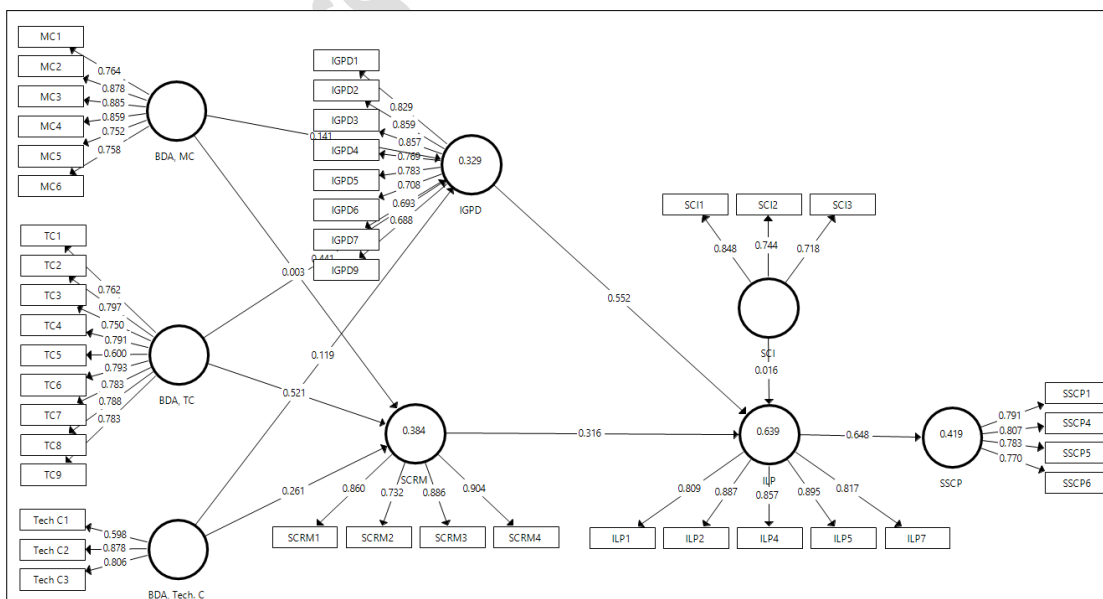
507 Insert Table 3 here

508  
 509 **4.4 Structural Model Assessment (SEM)**

510 A structural model was also assessed after the evaluation of the measurement model. For this  
 511 purpose, the significance of the model was assessed based on path coefficients, t-values, and  
 512 standard errors. The hypotheses were tested for the main and indirect effects through the  
 513 bootstrapping procedure in Smart PLS 3(Hair Jr, Sarstedt, Ringle, & Gudergan, 2017). Refer  
 514 to Table 4, Table 5, and Table 6, direct, indirect, and, moderating hypotheses were empirically  
 515 tested. Hypotheses were supported based on the critical ratio ( $t > 1.645$ ;  $P < 0.05$ ). It was found  
 516 that all the hypotheses are supported except few relationships.

517

518

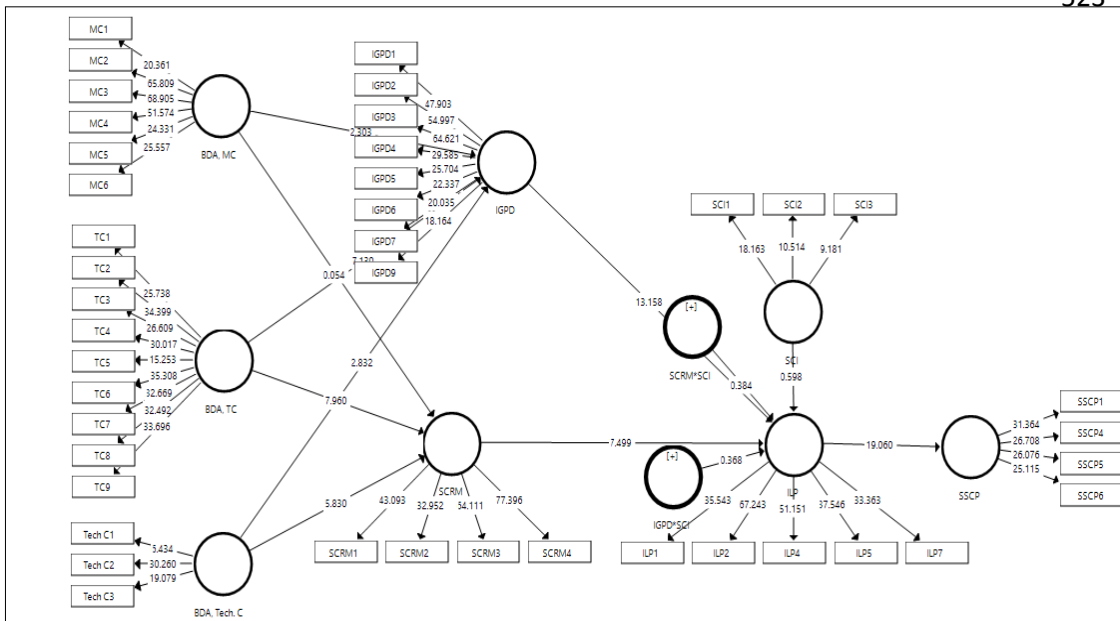


520 **Figure 2: Measurement Model Assessment**

521

522

523



538

539

540

541 **Figure 3: Structural Model Assessment**

542

543

544 Table 4

545 *Path Analysis*

546 Insert Table 4 here

547

548 Table 5

549 *Indirect Effects*

550 Insert Table 5 here

551 Table 6  
552 *Interaction Term*

553

554 Insert table 6 here

555

## 556 **5. Discussion**

557

558 This study discusses the importance of sustainable supply chain management; it is one of the  
559 most important firms' necessities. Sustainability is a major issue for organizations because  
560 every activity rotates around the organizations' operational activities. The current study aims  
561 to investigate how BDA plays a vital role in enhancing supply chain management's  
562 performance. However, this study also enhances the understanding of the implementation  
563 ways. The conceptual work is based on the novel concept involving BDA management, talent  
564 management, and technological capabilities to enrich how BDA enhances sustainable supply  
565 chain performance. The present model also discusses operational performance and supply  
566 chain risk management and shows that it will positively enhance supply chain performance  
567 sustainability. The empirical findings support the assumptions of the current study.

568 First, The learning capabilities and performance enhances the firm performance financially  
569 (Akgün et al., 2014). Similarly, the current study's findings also revealed that learning  
570 capabilities and performance also support the implementation of a sustainable supply chain.  
571 The study's findings explained that innovative product development enhances the learning  
572 capabilities and performance of the organization. On the other hand, BDA capabilities improve  
573 employee development and BDA talent management capability and have a significant  
574 relationship with the organization's innovative performance and learning performance (Aker  
575 et al., 2016). Moreover, this paper also adds the connection between talent management,  
576 employee management, and competitive performance by emphasizing sustainable supply chain  
577 management. Second, to achieve the benefits of BDA projects, many firms focus on supply  
578 chain innovativeness. Many firms used other factors in place of supply chain innovativeness;  
579 for that reason, they cannot get the full benefits from BDA projects (Marshall et al., 2015).  
580 Third, for gaining a competitive edge, BDA plays an important role and enhances the firm  
581 performance. Similarly, this study's assumptions are also supported by previous research  
582 (Dubey et al., 2019). Although, the linking between BDA has been improved the performance  
583 of the organizations and supply chain (Papadopoulos et al., 2017). However, operational  
584 performance sustainability is also improving through BDA.

585 Furthermore, Janssen et al. (2017) stated that the learning and innovation process leverage the  
586 BDA and big data would support sustainable operations. Previous work supports the  
587 assumption related to the talent management of BDA and the relationship with employee  
588 development (Marshall et al., 2015). Similarly, the supply chain innovativeness work as a  
589 moderator under this study also shows significant results. The result suggests that BDA  
590 management capabilities, talent management, and technological Management show sustainable  
591 outcomes.

592 The study results have presented evidence that the supply chain risk management results in the  
593 innovation and learning leading towards the sustainable supply chain performance. It is very  
594 crucial to manage the supply chain risks to ensure the organizational or industrial sustainability



595 in long-run. It is necessary to identify the supply chain risks so that the mitigation strategies  
596 can be applied to have the sustainability in long period of time (De Silva et al., 2021). The  
597 sustainable supply chain can be accomplished by enhancing the green knowledge resulting in  
598 the green innovations (Agyabeng-Mensah, Afum, & Baah, 2022). Moreover, sustainable  
599 supply chain can be accomplished by mitigation strategies for supply chain risks (Ali,  
600 Gongbing, & Mehreen, 2021) such as big data analytics, and innovativeness (Kalyar, Shafique,  
601 & Ahmad, 2019).

602

### 603 **5.1 Conclusion, future direction, and limitation**

604 The study attempted to examine the influence of big data analytics capabilities on the  
605 sustainable supply chain. In this regard, the study has considered the role of innovative green  
606 product development, supply chain risk management, and innovation and learning performance  
607 as mediators. Additionally, the supply chain innovativeness is considered as a moderator. The  
608 data were collected from multinational companies working in Pakistan. Based on the results of  
609 the study it is concluded that the big data analytics capabilities serve as an important factor that  
610 leads towards the development of sustainable supply chain performance. Organizations can  
611 accomplish sustainable supply chain performance by investing their resources in developing  
612 the big data analytical capabilities in different domains namely; management, talent, and  
613 technologies. Resultantly, this capabilities jointly serve as a dynamic capability resulting in  
614 sustainable supply chain performance. We conclude that BDA capabilities are essential for  
615 supply chain risk management. We propose that manufacturing firms are dependent on supply  
616 chain innovativeness to effectively manage innovation and learning performance to achieve  
617 sustainable supply chain management. We concluded that higher levels of supply chain  
618 innovativeness reduce the supply chain risks, which implies greater learning and innovation is  
619 more dependent on supply chain innovativeness.

620 Our results provide significant evidence; supply chain innovativeness appears a key signal for  
621 the firms to generate a high level of innovation and learning performance while reducing the  
622 supply risk management. According to the research design, a few limitations must consider by  
623 future researchers. The primary rule is the utilization of cross-sectional data. Secondly, only  
624 multinational companies are considered under this study, which provides regulations in  
625 outcomes, and future research, drawing from more extensive areas may reduce the limitation.  
626 Third, the research group grounded the hypothetical system utilizing dynamic capacity see  
627 hypothesis. Future research can improve by working with these outcomes as a base.  
628 Additionally, create complex models by adding other variables which affect green product  
629 development. Aligned with the previous study suggestions it is suggested that future studies  
630 should examine the leadership role towards creating sustainable development performance  
631 outcomes (Awan & Khan, 2021).

632

### 633 **5.2. Theoretical Implications**

634

635 From a theoretical perspective, the present study has several implications which are as follows;  
636 this study is the first to investigate the moderating impact of supply chain innovativeness  
637 between the relationship between supply chain risks and innovation learning and performance.  
638 Despite the increasing interest in green or sustainable supply chain management, little research

639 exists on how firms their supply chain risk management and innovative product development  
640 in the presence of the BDA capabilities. Our study addresses this research gap by developing  
641 a conceptual framework in the post-COVID-19 in an emerging country context. Additionally,  
642 the study has advanced the literature on sustainable supply chain performance and its  
643 indicators. More precisely, it has presented evidence on of how big data analytics capabilities  
644 can result in sustainable supply chain performance through mediators as process explainers.  
645 Additionally, it is one of the few studies that advanced the literature on the big data analytical  
646 capabilities and sustainable supply chain performance while considering the boundary  
647 condition of supply chain innovativeness. It advanced the literature by considering the  
648 innovations in supply chain and risk management from the supply chain perspective.

649

### 650 **5.3. Managerial implications**

651

652 Our outcomes provide direction to consultants and managers in post-COVID-19 situations who  
653 are engaged in implementing BDA in firms. The findings present a few managerial associations  
654 seeking to accomplish operational distinction, especially in multinational companies.  
655 Managers of the organizations can achieve operational qualities from the significant  
656 relationship between BDA, the management capacity, and innovative green product  
657 development in the post-COVID-19 situation. Specifically, BDA has exhaustively changed the  
658 way of working of every association. Moreover, the entry of BDA has empowered associations  
659 to deal with voluminous information in a compelling and developed form. The study results  
660 recommend that effective utilization of big data analytics can bolster sustainable products in  
661 operation and organizational supply chain functions, providing help to fully utilize BDA.  
662 Sustainability must be emphasized extensively regardless of the organizational operational  
663 domains since it enables the organizations to maintain their corporate sustainability in the  
664 ecosystem (Kanwal & Awan, 2020).

665 Managers of the organizations can improve their resource development and redevelopment  
666 related to supply chains by incorporating the big data analytical capabilities such as managerial,  
667 talent, and technological respectively as the larger data sets will enable them to identify the  
668 previous trends and also forecast the future trends related to the supply chain, ultimately,  
669 leading towards the higher performance. Additionally, they can also seek resources  
670 development from the supplier perspectives as the larger data sets will enable them to see the  
671 shortages in the supply chain and also look for the future needs so they can timely take an  
672 action. The study has presented the findings from the dynamic capability perspective. So the  
673 managers must emphasize developing the organizational dynamic capabilities that will enable  
674 them to explore and exploit the opportunities resulting in sustainable supply chain performance.  
675 The managers are required to alter the existing supply chains by not only considering the  
676 financial aspects but environmental aspects (Rahman et al., 2020) and social aspects (Munny  
677 et al., 2019) as well for creating a sustainable supply chain management (Tumpa et al., 2019).

678

679

680

681

682

683 **Declarations**

684

- 685 • **Ethical Approval:** Not Applicable
- 686 • **Consent to Participate:** Consent to participate in the study was obtained from the  
687 respondents.
- 688 • **Consent to Publish:** All authors are agreed to publish this manuscript.
- 689 • **Funding:** No funding was received to assist with the preparation of this manuscript.
- 690 • **Competing Interests:** The authors declare that there is no competing interest.
- 691 • **Availability of data and materials:** Data and relevant materials are available on  
692 request to corresponding authors.

693

694 **References:**

695

- 696 Agyabeng-Mensah, Y., Afum, E., & Baah, C. (2022). Green corporate reputation and  
697 innovation: the role of non-supply chain learning and green supply chain knowledge.  
698 *International Journal of Emerging Markets*.
- 699 Ahmad, M. A., Iteng, R., Saad, R., & Abd Rahim, M. K. I. (2018). The criteria of sustainable  
700 product development and organizational performance. *International Journal of Supply  
701 Chain Management (IJSCM)*, 7(5), 497-501.
- 702 Akgün, A. E., Ince, H., Imamoglu, S. Z., Keskin, H., & Kocoglu, İ. (2014). The mediator role  
703 of learning capability and business innovativeness between total quality management  
704 and financial performance. *International Journal of Production Research*, 52(3), 888-  
705 901.
- 706 Akter, S., Wamba, S. F., Gunasekaran, A., Dubey, R., & Childe, S. J. (2016). How to improve  
707 firm performance using big data analytics capability and business strategy alignment?  
708 *International Journal of Production Economics*, 182, 113-131.
- 709 Alam, S. T., Ahmed, S., Ali, S. M., Sarker, S., & Kabir, G. (2021). Challenges to COVID-19  
710 vaccine supply chain: Implications for sustainable development goals. *International  
711 Journal of Production Economics*, 239, 108193.
- 712 Ali, F., Rasoolimanesh, S. M., Sarstedt, M., Ringle, C. M., & Ryu, K. (2018). An assessment  
713 of the use of partial least squares structural equation modeling (PLS-SEM) in  
714 hospitality research. *International Journal of Contemporary Hospitality Management*.
- 715 Ali, S., Poulouva, P., Yasmin, F., Danish, M., Akhtar, W., & Usama Javed, H. M. (2020). How  
716 big data analytics boosts organizational performance: The mediating role of the  
717 sustainable product development. *Journal of Open Innovation: Technology, Market,  
718 and Complexity*, 6(4), 190.
- 719 Ali, Z., Gongbing, B., & Mehreen, A. (2021). Do vulnerability mitigation strategies influence  
720 firm performance: the mediating role of supply chain risk. *International Journal of  
721 Emerging Markets*.
- 722 Awan, U. (2019). Impact of social supply chain practices on social sustainability performance  
723 in manufacturing firms. *International Journal of Innovation and Sustainable  
724 Development*, 13(2), 198-219.
- 725 Awan, U., Bhatti, S. H., Shamim, S., Khan, Z., Akhtar, P., & Balta, M. E. (2021). The Role of  
726 Big Data Analytics in Manufacturing Agility and Performance: Moderation–Mediation  
727 Analysis of Organizational Creativity and of the Involvement of Customers as Data  
728 Analysts. *British Journal of Management*.
- 729 Awan, U., & Khan, S. A. R. (2021). Mediating role of sustainable leadership in buyer-supplier  
730 relationships: An supply chain performance: An empirical study. *LogForum*, 17(1).

- 731 Bag, S. (2017). Big data and predictive analysis is key to superior supply chain performance:  
732 a South African experience. *International Journal of Information Systems and Supply*  
733 *Chain Management (IJSSCM)*, 10(2), 66-84.
- 734 Bag, S., Wood, L. C., Xu, L., Dhamija, P., & Kayikci, Y. (2020). Big data analytics as an  
735 operational excellence approach to enhance sustainable supply chain performance.  
736 *Resources, Conservation and Recycling*, 153, 104559.
- 737 Begum, S., Ashfaq, M., Xia, E., & Awan, U. (2022). Does green transformational leadership  
738 lead to green innovation? The role of green thinking and creative process engagement.  
739 *Business Strategy and the Environment*, 31(1), 580-597.
- 740 Bleadly, A., Ali, A. H., & Ibrahim, S. B. (2018). Dynamic capabilities theory: pinning down a  
741 shifting concept. *Academy of Accounting and Financial Studies Journal*, 22(2), 1-16.
- 742 Chakphet, T., Saenpakdee, M., Pongsiri, T., & Jernsittiparsert, K. (2020). The Role of Big  
743 Data Analytics in the Relationship among the Collaboration Types, Supply Chain  
744 Management and Market Performance of Thai Manufacturing Firms. *International*  
745 *Journal of Supply Chain Management*, 9, 28-36.
- 746 Choi, T.-M. (2020). Innovative “bring-service-near-your-home” operations under Corona-  
747 virus (COVID-19/SARS-CoV-2) outbreak: Can logistics become the messiah?  
748 *Transportation Research Part E: Logistics and Transportation Review*, 140, 101961.
- 749 Chowdhury, N. A., Ali, S. M., Paul, S. K., Mahtab, Z., & Kabir, G. (2020). A hierarchical  
750 model for critical success factors in apparel supply chain. *Business Process*  
751 *Management Journal*.
- 752 Dangelico, R. M., Pujari, D., & Pontrandolfo, P. (2017). Green product innovation in  
753 manufacturing firms: A sustainability-oriented dynamic capability perspective.  
754 *Business Strategy and the Environment*, 26(4), 490-506.
- 755 De Silva, U. S. K., Paul, A., Hasan, K. W., Paul, S. K., Ali, S. M., & Chakraborty, R. K.  
756 (2021). Examining risks and strategies for the spice processing supply chain in the  
757 context of an emerging economy. *International Journal of Emerging Markets*.
- 758 Dubey, R., Gunasekaran, A., & Childe, S. J. (2019). Big data analytics capability in supply  
759 chain agility. *Management Decision*.
- 760 Gunasekaran, A., Papadopoulos, T., Dubey, R., Wamba, S. F., Childe, S. J., Hazen, B., &  
761 Akter, S. (2017). Big data and predictive analytics for supply chain and organizational  
762 performance. *Journal of Business Research*, 70, 308-317.
- 763 Hair, J. F., Black, W. C., Babin, B. J., & Anderson, R. E. (2010). Confirmatory factor analysis.  
764 In *Multivariate Data Analysis*.
- 765 Hair Jr, J. F., Hult, G. T. M., Ringle, C. M., & Sarstedt, M. (2021). *A primer on partial least*  
766 *squares structural equation modeling (PLS-SEM)*: Sage publications.
- 767 Hall, J. (2000). Environmental supply chain dynamics. *Journal of cleaner production*, 8(6),  
768 455-471.
- 769 Hammervoll, T., Jensen, L.-M., & Beske, P. (2012). Dynamic capabilities and sustainable  
770 supply chain management. *International journal of physical distribution & logistics*  
771 *management*.
- 772 Hult, G. T. M., Hurley, R. F., & Knight, G. A. (2004). Innovativeness: Its antecedents and  
773 impact on business performance. *Industrial marketing management*, 33(5), 429-438.
- 774 Ilyas, S., Hu, Z., & Wiwattanakornwong, K. (2020). Unleashing the role of top management  
775 and government support in green supply chain management and sustainable  
776 development goals. *Environmental Science and Pollution Research*, 27(8), 8210-8223.
- 777 Ivanov, D. (2020). Predicting the impacts of epidemic outbreaks on global supply chains: A  
778 simulation-based analysis on the coronavirus outbreak (COVID-19/SARS-CoV-2)  
779 case. *Transportation Research Part E: Logistics and Transportation Review*, 136,  
780 101922.

- 781 Janssen, M., van der Voort, H., & Wahyudi, A. (2017). Factors influencing big data decision-  
782 making quality. *Journal of Business Research*, 70, 338-345.
- 783 Kalyar, M. N., Shafique, I., & Ahmad, B. (2019). Effect of innovativeness on supply chain  
784 integration and performance: Investigating the moderating role of environmental  
785 uncertainty. *International Journal of Emerging Markets*.
- 786 Kamble, S. S., & Gunasekaran, A. (2020). Big data-driven supply chain performance  
787 measurement system: a review and framework for implementation. *International*  
788 *Journal of Production Research*, 58(1), 65-86.
- 789 Kanwal, N., & Awan, U. (2020). Role of design thinking and biomimicry in leveraging  
790 sustainable innovation.
- 791 Karmaker, C. L., Ahmed, T., Ahmed, S., Ali, S. M., Moktadir, M. A., & Kabir, G. (2021).  
792 Improving supply chain sustainability in the context of COVID-19 pandemic in an  
793 emerging economy: Exploring drivers using an integrated model. *Sustainable*  
794 *production and consumption*, 26, 411-427.
- 795 Karuppiah, K., Sankaranarayanan, B., Ali, S. M., & Paul, S. K. (2021). Key challenges to  
796 sustainable humanitarian supply chains: lessons from the covid-19 pandemic.  
797 *Sustainability*, 13(11), 5850.
- 798 Khan, S. A. R., & Qianli, D. (2017). Impact of green supply chain management practices on  
799 firms' performance: an empirical study from the perspective of Pakistan.  
800 *Environmental Science and Pollution Research*, 24(20), 16829-16844.
- 801 Latif, Z., Tunio, M. Z., Pathan, Z. H., Jianqiu, Z., Ximei, L., & Sadozai, S. K. (2018). *A review*  
802 *of policies concerning development of big data industry in Pakistan: Subtitle:*  
803 *Development of big data industry in Pakistan*. Paper presented at the 2018 international  
804 conference on computing, mathematics and engineering technologies (iCoMET).
- 805 Liao, R., & Fan, Z. (2020). *Supply chains have been upended. Here's how to make them more*  
806 *resilient*. Paper presented at the World Economic Forum.
- 807 Mahesar, H. A., Chaudhry, N. I., & Tariq, U. (2017). INTEGRATING CUSTOMER  
808 RELATIONSHIP MANAGEMENT WITH BIG DATA ANALYTICS IN RETAIL  
809 STORES: A CASE OF HYPER-STAR AND METRO. *Journal of Business Strategies*,  
810 11(2), 141-158.
- 811 Mandal, S. (2019). The influence of big data analytics management capabilities on supply chain  
812 preparedness, alertness and agility: an empirical investigation. *Information Technology*  
813 *& People*.
- 814 Marshall, A., Mueck, S., & Shockley, R. (2015). How leading organizations use big data and  
815 analytics to innovate. *Strategy & Leadership*.
- 816 Mathivathanan, D., Kannan, D., & Haq, A. N. (2018). Sustainable supply chain management  
817 practices in Indian automotive industry: A multi-stakeholder view. *Resources,*  
818 *Conservation and Recycling*, 128, 284-305.
- 819 Mishra, D., Gunasekaran, A., Papadopoulos, T., & Childe, S. J. (2018). Big Data and supply  
820 chain management: a review and bibliometric analysis. *Annals of Operations Research*,  
821 270(1-2), 313-336.
- 822 Moktadir, M. A., Ali, S. M., Rajesh, R., & Paul, S. K. (2018). Modeling the interrelationships  
823 among barriers to sustainable supply chain management in leather industry. *Journal of*  
824 *Cleaner Production*, 181, 631-651.
- 825 Mukhtar, M. A., Baloch, N. A., & Khattak, S. R. (2019). Dynamic Capability & Firm  
826 Performance: Mediating Role Of Learning Orientation, Organizational Culture &  
827 Corporate Entrepreneurship: A Case Study Of Sme's Of Pakistan. *J Manag Sci*, 13,  
828 119-128.

829 Munny, A. A., Ali, S. M., Kabir, G., Moktadir, M. A., Rahman, T., & Mahtab, Z. (2019).  
830 Enablers of social sustainability in the supply chain: An example of footwear industry  
831 from an emerging economy. *Sustainable Production and Consumption*, 20, 230-242.  
832 Oke, A. E., Ogunsami, D. R., & Ogunlana, S. (2012). Establishing a common ground for the  
833 use of structural equation modelling for construction related research studies.  
834 *Construction economics and building*, 12(3), 89-94.  
835 Papadopoulos, T., Gunasekaran, A., Dubey, R., Altay, N., Childe, S. J., & Fosso-Wamba, S.  
836 (2017). The role of Big Data in explaining disaster resilience in supply chains for  
837 sustainability. *Journal of cleaner production*, 142, 1108-1118.  
838 Paul, S., Ali, S. M., Hasan, M. A., Paul, S. K., & Kabir, G. (2022). Critical Success Factors for  
839 Supply Chain Sustainability in the Wood Industry: An Integrated PCA-ISM Model.  
840 *Sustainability*, 14(3), 1863.  
841 Pervan, M., Curak, M., & Pavic Kramaric, T. (2017). The influence of industry characteristics  
842 and dynamic capabilities on firms' profitability. *International Journal of Financial*  
843 *Studies*, 6(1), 4.  
844 Pujari, D., Wright, G., & Peattie, K. (2003). Green and competitive: Influences on  
845 environmental new product development performance. *Journal of Business Research*,  
846 56(8), 657-671.  
847 Qwabe, A. (2020). *Big data analytics as a dynamic capability to enhance customer value*  
848 *management for sustainable competitive advantage*. University of Pretoria,  
849 Rahman, T., Ali, S. M., Moktadir, M. A., & Kusi-Sarpong, S. (2020). Evaluating barriers to  
850 implementing green supply chain management: An example from an emerging  
851 economy. *Production Planning & Control*, 31(8), 673-698.  
852 Raian, S., Ali, S. M., Sarker, M. R., Sankaranarayanan, B., Kabir, G., Paul, S. K., &  
853 Chakraborty, R. K. (2022). Assessing sustainability risks in the supply chain of the  
854 textile industry under uncertainty. *Resources, Conservation and Recycling*, 177,  
855 105975.  
856 Salehan, M., & Kim, D. J. (2016). Predicting the performance of online consumer reviews: A  
857 sentiment mining approach to big data analytics. *Decision Support Systems*, 81, 30-40.  
858 Sezen, B. (2008). Relative effects of design, integration and information sharing on supply  
859 chain performance. *Supply chain management: An international journal*.  
860 Shah, N., Irani, Z., & Sharif, A. M. (2017). Big data in an HR context: Exploring organizational  
861 change readiness, employee attitudes and behaviors. *Journal of Business Research*, 70,  
862 366-378.  
863 Shahed, K. S., Azeem, A., Ali, S. M., & Moktadir, M. (2021). A supply chain disruption risk  
864 mitigation model to manage COVID-19 pandemic risk. *Environmental Science and*  
865 *Pollution Research*, 1-16.  
866 Shamim, S., Zeng, J., Shariq, S. M., & Khan, Z. (2019). Role of big data management in  
867 enhancing big data decision-making capability and quality among Chinese firms: A  
868 dynamic capabilities view. *Information & Management*, 56(6), 103135.  
869 Shohan, S., Ali, S., Kabir, G., Ahmed, S. K., Suhi, S., & Haque, T. (2019). Green supply chain  
870 management in the chemical industry: structural framework of drivers. *International*  
871 *Journal of Sustainable Development & World Ecology*, 26(8), 752-768.  
872 Shukor, A. A. A., Newaz, M. S., Rahman, M. K., & Taha, A. Z. (2020). Supply chain  
873 integration and its impact on supply chain agility and organizational flexibility in  
874 manufacturing firms. *International Journal of Emerging Markets*.  
875 Siddique, M. N. A., Hasan, K. W., Ali, S. M., Moktadir, M. A., Paul, S., & Kabir, G. (2021).  
876 Modeling drivers to big data analytics in supply chains. *Journal of Production Systems*  
877 *and Manufacturing Science*.

- 878 Silvestre, B. S. (2015). Sustainable supply chain management in emerging economies:  
879 Environmental turbulence, institutional voids and sustainability trajectories.  
880 *International Journal of Production Economics*, 167, 156-169.
- 881 Singh, N. P., & Singh, S. (2019). Building supply chain risk resilience. *Benchmarking: An*  
882 *International Journal*.
- 883 Singh, S. K., & El-Kassar, A.-N. (2019). Role of big data analytics in developing sustainable  
884 capabilities. *Journal of cleaner production*, 213, 1264-1273.
- 885 Sivarajah, U., Kamal, M. M., Irani, Z., & Weerakkody, V. (2017). Critical analysis of Big Data  
886 challenges and analytical methods. *Journal of Business Research*, 70, 263-286.
- 887 Tan, K. H., & Zhan, Y. (2017). Improving new product development using big data: a case  
888 study of an electronics company. *R&D Management*, 47(4), 570-582.
- 889 Teece, D. J. (2016). Dynamic capabilities and entrepreneurial management in large  
890 organizations: Toward a theory of the (entrepreneurial) firm. *European Economic*  
891 *Review*, 86, 202-216.
- 892 Tiwari, S., Wee, H. M., & Daryanto, Y. (2018). Big data analytics in supply chain management  
893 between 2010 and 2016: Insights to industries. *Computers & Industrial Engineering*,  
894 115, 319-330.
- 895 Trabucchi, D., & Buganza, T. (2018). Data-driven innovation: Switching the perspective on  
896 Big Data. *European Journal of Innovation Management*.
- 897 Tumpa, T. J., Ali, S. M., Rahman, M. H., Paul, S. K., Chowdhury, P., & Khan, S. A. R. (2019).  
898 Barriers to green supply chain management: An emerging economy context. *Journal of*  
899 *Cleaner Production*, 236, 117617.
- 900 United Nations. (2022). Build resilient infrastructure, promote inclusive and sustainable  
901 industrialization and foster innovation. Retrieved from <https://sdgs.un.org/goals/goal9>
- 902 Wamba, S. F., Gunasekaran, A., Akter, S., Ren, S. J.-f., Dubey, R., & Childe, S. J. (2017). Big  
903 data analytics and firm performance: Effects of dynamic capabilities. *Journal of*  
904 *Business Research*, 70, 356-365.
- 905 Wang, G., Gunasekaran, A., Ngai, E. W., & Papadopoulos, T. (2016). Big data analytics in  
906 logistics and supply chain management: Certain investigations for research and  
907 applications. *International Journal of Production Economics*, 176, 98-110.
- 908 Yeoman, R., & Santos, M. M. (2019). A complex systems model for transformative supply  
909 chains in emerging markets. *International Journal of Emerging Markets*.
- 910 Zhan, Y., Tan, K. H., Li, Y., & Tse, Y. K. (2018). Unlocking the power of big data in new  
911 product development. *Annals of Operations Research*, 270(1-2), 577-595.
- 912 Zhao, R., Liu, Y., Zhang, N., & Huang, T. (2017). An optimization model for green supply  
913 chain management by using a big data analytic approach. *Journal of cleaner*  
914 *production*, 142, 1085-1097.
- 915 Zhong, R. Y., Newman, S. T., Huang, G. Q., & Lan, S. (2016). Big Data for supply chain  
916 management in the service and manufacturing sectors: Challenges, opportunities, and  
917 future perspectives. *Computers & Industrial Engineering*, 101, 572-591.

919

920 Table 4

921 *Path Analysis*

	<b>Relationships</b>	<b>Std. beta</b>	<b>Std. Error</b>	<b>t-Value</b>	<b>L.L</b>	<b>U.L.</b>	<b>R<sup>2</sup></b>	<b>VIF</b>	<b>Decision</b>
H1	BDA, MC -> IGPD	0.141	0.061	2.320	0.044	0.244	0.329	1.946	Supported <sup>924</sup>
H2	BDA, TC -> IGPD	0.441	0.058	7.572	0.343	0.539		1.995	Supported <sup>925</sup>
H3	BDA, Tech. C -> IGPD	0.119	0.044	2.691	0.039	0.184		1.063	Supported <sup>926</sup>
H4	BDA, MC -> SCRM	0.003	0.056	0.051	-0.089	0.094	0.384	1.946	Not Supported <sup>927</sup>
H5	BDA, TC -> SCRM	0.521	0.063	8.329	0.421	0.628		1.995	Supported <sup>928</sup>
H6	BDA, Tech. C -> SCRM	0.261	0.043	6.055	0.187	0.322		1.063	Supported <sup>929</sup>
H7	IGPD -> ILP	0.552	0.043	12.901	0.478	0.624	0.639	1.811	Supported <sup>930</sup>
H11	SCRM -> ILP	0.316	0.043	7.349	0.248	0.386		1.081	Supported <sup>931</sup>
H15	ILP -> SSCP	0.648	0.032	20.207	0.583	0.69	0.419	1.001	Supported <sup>932</sup>

	<b>Relationships</b>	<b>Std. beta</b>	<b>Std. Error</b>	<b>t-Value</b>	<b>L.L</b>	<b>U.L.</b>	<b>Decision</b>
H8	BDA, MC -> IGPD -> ILP	0.078	0.035	2.244	0.025	0.135	Supported
H9	BDA, TC -> IGPD -> ILP	0.243	0.035	6.884	0.183	0.299	Supported
H10	BDA, Tech. C -> IGPD -> ILP	0.065	0.025	2.64	0.023	0.102	Supported
H12	BDA, MC -> SCRM -> ILP	0.001	0.018	0.051	-0.028	0.032	Not Supported
H13	BDA, TC -> SCRM -> ILP	0.165	0.032	5.112	0.118	0.223	Supported
H14	BDA, Tech. C -> SCRM -> ILP	0.082	0.017	4.899	0.058	0.114	Supported
H16	IGPD -> ILP -> SSCP	0.358	0.031	11.664	0.306	0.408	Supported

Table 5  
*Indirect Effects*



H17    SCRM -> ILP -> SSCP                    0.205    0.033    6.257    0.152    0.26    Supported

934

---

935  
936  
937  
938  
939  
940  
941

Authors submitted version

942 Table 6

943 *Interaction Term*

	<b>Relationships</b>	<b>Std. beta</b>	<b>Std. Error</b>	<b>t- Value</b>	<b>L.L</b>	<b>U.L.</b>	<b>Decision</b>
H18	IGPD*SCI -> ILP	0.117	0.040	2.925	0.168	0.257	Supported
H19	SCRM*SCI -> ILP	-0.031	-0.035	0.885	-0.084	0.109	Not Supported

944

945

946

947

948

949

950

951

952

953

954

955

956

957

Authors submitted version