

1 **A scientometric analysis and critical review of digital twin applications in project operation**
2 **and maintenance**

3 **Abstract**

4 **Purpose** – Recent emerging information technologies like digital twin (DT) provide new
5 concepts and transform information management processes in the architecture, engineering, and
6 construction (AEC) industries. Although numerous articles are pertinent to DT applications,
7 existing research areas and potential directions related to the state-of-the-art DT in project
8 operation and maintenance (O&M) are yet to be studied. Therefore, this paper aims to review
9 the state-of-the-art research on DT applications in project O&M.

10 **Design/methodology/approach** – To do this, the current review adopted four methodological
11 steps, including literature search, literature selection, science mapping analysis, and qualitative
12 discussion to gain a deeper understanding of DT in project O&M. The impact and contribution
13 of keywords and documents were examined from a total of 444 journal articles retrieved from
14 the Scopus database.

15 **Findings** – Five mainstream research topics were identified, including (1) DT-based artificial
16 intelligence (AI) technology for project O&M, (2) DT-enabled smart city and sustainability, (3)
17 DT applications for project asset management, (4) Blockchain-integrated DT for project O&M,
18 and (5) DT for advanced project management. Subsequently, research gaps and future research
19 directions were proposed.

20 **Originality** – This study intends to raise awareness of future research by summarising the
21 current DT development phases and their impact on DT implementation in project O&M among
22 researchers and practitioners.

23 **Keywords:** Digital twin; Information technologies; Project operation and maintenance;
24 Scientometric; Construction industry

25 **Paper type:** Literature review

26 **1. Introduction**

27 Data has become one of the most valuable assets in the world today; the data from engineering
28 assets and systems are essential for equipment monitoring and diagnostics (Shi et al., 2023). The
29 use of data also significantly influences decision-making across the project management life cycle
30 because the capability and quality of project performance have increased due to enterprises using
31 descriptive project data analytics (Kaewunruen and Lian, 2019). The development of simulation
32 and data-relevant technologies is growing within the Fourth Industrial Revolution (4IR). Deloitte
33 Insight (2021) suggested that over 70% of respondents to the investigation expressed that their
34 organisations are utilising or exploiting artificial intelligence (AI). Analytics and cognition will
35 have the second-largest demonstrable influence on enterprises in the next three years, according
36 to more than 90% of chief information officers (CIOs) and senior technology directors (Deloitte
37 Insight, 2020). Therefore, the construction industry has focused primarily on smart project asset
38 management under the 4IR era.

39 Project operation and maintenance (O&M) typically lasted more than 50 years, making it difficult
40 to challenge smart construction management (Wang et al., 2023). Each project's construction costs
41 include a sizeable portion of the O&M phase. Data information is essential for building daily
42 management and entity equipment such as cameras. One of the biggest challenges in the process
43 of O&M management is maintaining the integrity, validity, and interactivity of data (Wetzel and
44 Thabet, 2015). The evidence shows that the construction industry can contribute to nearly 10% of
45 economics in different countries; the digital twin (DT) has been utilized in many regions and
46 countries such as China, Australia, the United States, and the United Kingdom (Opoku et al., 2021).
47 Hence, the value of DT in the construction industries' project O&M plays a significant role in
48 future sector development, occupational optimization, and work process improvement (Feng et al.,
49 2023).

50 Meanwhile, DT as the problem detection and decision-making tool has been proven to facilitate
51 data utilization by integrating it with other technologies' data for daily project O&M (Müller-
52 Zhang et al., 2023). As such, DT applications can enhance collaboration through real-time
53 information sharing in many industries, such as architecture, manufacturing, engineering, and
54 construction (Lu et al., 2020a); (Dabirian et al., 2023). In the traditional approach, designers adopt
55 computer simulation and engineering tools to calculate and design project life cycles and physical
56 detecting mechanisms (Chen et al., 2022). They try to optimise the procedure through accurate

57 calculation to save cost, but this method lacks consideration for strategy limitation and the
58 relationship of applicants' configuration. As the computer industry and AI evolved (Gao et al.,
59 2023), DT—a form of improved algorithms and cutting-edge computer technologies—has made
60 real-time monitoring and digital power conceivable. DT can present every physical object, process,
61 and system. It provides a dashboard that can monitor past and present operations and predict future
62 actions by combining software analysis, AI, and machine learning (ML) data, then update any
63 changes in the physical environment (Michie et al., 2017).

64 DT technology has been applied in several sectors. For example, Xiong et al. (2021) mentioned
65 that NASA was the first organization that applied DT technology to continuously monitor
66 spacecraft status to prevent degradation and failure in 2002. General Electric's (GE) digital
67 department has begun the exploration of a jet engine that can predict the commercial outcome and
68 components' life. Jiang (2021) argued that the service application layer in the field of construction
69 can also display the O&M status of construction lines on a variety of platforms using modules for
70 construction quality presentation, building process control, change management, work progress
71 feedback, device failure diagnosis, health status testing, and O&M. However, a precise knowledge
72 of how to deal with the future direction of integrating DT with current technologies and systems
73 is lacking, as well as comprehensive DT adaption plans (Zhao et al., 2022); (Grüner et al., 2023).
74 Since the architecture, engineering, and construction (AEC) industries are undergoing a
75 burgeoning digital transformation, data virtualisation technologies and representation levels have
76 become a new and critical research direction for these industries (D'Urso et al., 2024). Opoku et
77 al. (2021) analyzed the value of DT in six application areas such as facility management, logistics
78 management, monitoring and control, and structural interaction in the project lifecycle. These six
79 segments belong to specific branches of the project O&M, so it proves the importance of DT for
80 upgrading the operational model of the construction industry. As noted by Boje et al. (2020),
81 building information modeling (BIM) provides the protocols for data standards and monitoring
82 activities for sensor networks to increase the added value of equipment data, while DT technology
83 makes use of the synchronization of the bi-directional cyber-physical data flows to reduce the
84 BIM's control capability gap.

85 Existing studies have focused on the benefits and DT integration with other technologies for
86 organizational performance, facilitating the maturity of digital transformation (Broo and Schooling,
87 2023). However, the identification or prediction of faults and real-time monitoring of machine

88 equipment operating conditions represent the biggest obstacles to complete automation of
89 machinery and improve project O&M (Deebak and Al-Turjman, 2022). Even though the quality
90 of project O&M and diverse process flow control can benefit from its application (Zhang et al.,
91 2024), it would not be realized if the basic problems cannot be solved and relevant changes are not
92 appropriately accomplished. Examples include interoperability and standards of data processing
93 within different technologies (Ramonell et al., 2023), challenges of data collection and analyzes
94 for supporting decision-making (Kamari and Ham, 2022), and practical innovations in project
95 O&M procedures for facility managers or O&M managers to enable real-time monitoring and
96 service-based production (Müller-Zhang et al., 2023). The challenges of DT applications in project
97 O&M necessitate its alignment with the organization's strategy and social acceptance which
98 conform to the requirements of digital transformation of project management. Therefore,
99 understanding how to reap the rewards of DT deployment is more crucial than understanding why
100 to use this technology (Love and Matthews, 2019). Several researchers have focused on DT
101 applications for smart construction and carbon emissions in building projects (Yevu et al., 2023),
102 safety management (Agnusdei et al., 2021), building construction industry (Long et al., 2024), and
103 smart buildings (Ghansah and Lu, 2024). Despite previous review efforts, there is limited research
104 related to the application of DT in project O&M. As a result, this study explores the adoption of
105 DT applications in project O&M phase of construction lifecycle processes, and provides research
106 gaps and future research directions that are beneficial to researchers and practitioners and for
107 advancing research in this field.

108 Given the above, this study aims to conduct a Scientometric and critical review of published
109 articles in the Scopus database related to DT applications in project O&M in the last 10 years (i.e.,
110 from 2014 to 2024). Specific research questions that were formulated to achieve the stated aim
111 include:

- 112 1) What are the annual research publication trends and relevant peer-reviewed journals on
113 DT in project O&M?
- 114 2) What are the scientometric analyses on co-occurrence keywords and documents?
- 115 3) What are the mainstream topics identified by DT in project O&M?
- 116 4) What are the future research directions on DT in project O&M?

117 The results of this review could assist researchers, policymakers, and practitioners to enhance the
118 understanding of recent developments and future demands of DT application in project O&M, and

119 how it contributes to the digital transformation of project management. Likewise, the findings can
120 help other researchers to advance potential research directions for DT integration with emerging
121 digital technologies such as AI, blockchain, and IoT, which would facilitate decision-making, fault
122 diagnosis and forecasting in the process of project O&M. Besides, this review study would draw
123 the attention of policymakers and practitioners to the importance of data/information management
124 to enable the application of DT in complex project management scenarios.

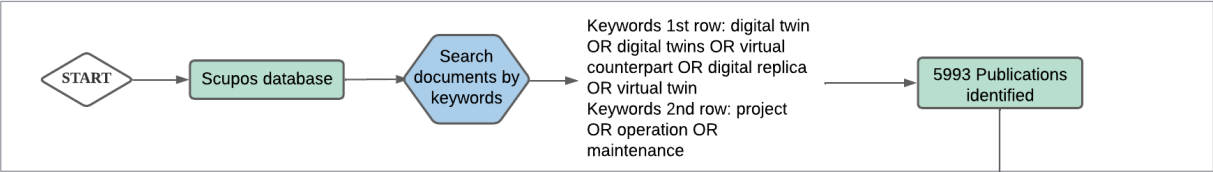
125 The remainder of the review paper is as follows. Section 2 elaborates on the research methodology.
126 The results of the annual publication, relevant peer-reviewed journals, and scientometric analyses
127 are reported in Section 3. Discussions of mainstream research topics, research gaps, and future
128 research directions are provided in Section 4. Section 5 summarises the conclusions of this review
129 paper, while Section 6 highlights its limitations and future research directions.

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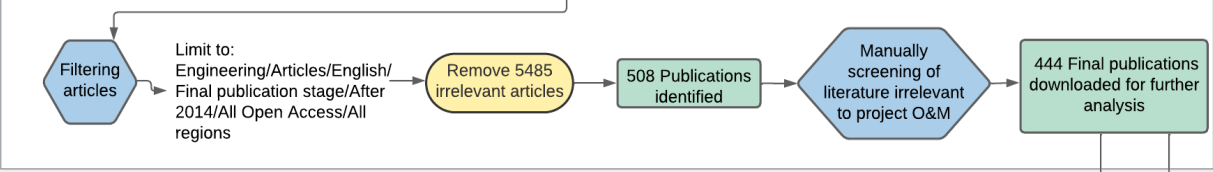
131 **2. Research methodology**

132 This study adopted a scientometric analysis and critical review method to analyse and virtualise
133 related articles on DT applications in project O&M. An in-depth understanding of the structure and
134 research trends of the studied domain was represented by maps or charts. The Scopus database was
135 used to search for relevant publications and served as a source of data collection. A systematic
136 review was conducted to synthesize the recent and existing research studies. The findings from the
137 critical review would be beneficial for professional knowledge, theories, and promote
138 understanding of research trends. An overview of this review research process is illustrated in Fig.1.

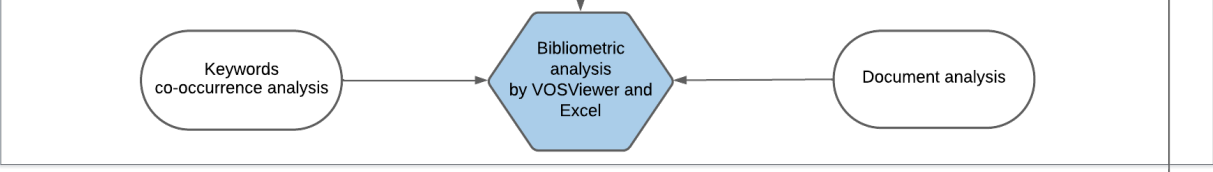
Stage 1 Literature search



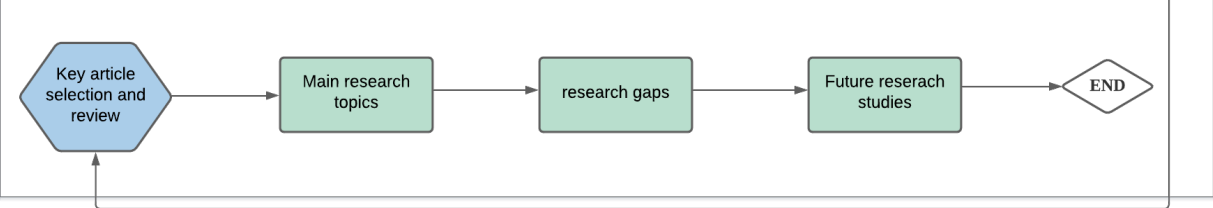
Stage 2 Literature selection



Stage 3 Science mapping analysis



Stage 4 Qualitative discussion



139
140 **Fig. 1.** An overview of the research process

141
142 *2.1. Stage 1 Literature search*

143 The initial step is to search for relevant publications in the Scopus database that would assist in
 144 presenting the results for the annual publication trends and relevant peer-reviewed journals on DT
 145 in project O&M. Scopus and Web of Science are the most scientific literature search databases.
 146 However, Scopus database covers more broader range of multiple disciplines of journals and
 147 articles compared to the Web of Science (Mongeon and Paul-Hus, 2016). Given the
 148 interdisciplinary nature of DT applications in project O&M, the Scopus database was selected to
 149 capture a wider range of perspectives or diverse disciplines that contribute to the understanding of
 150 this study. Furthermore, Scopus has more citation counts and is recognized as performing better
 151 than the Web of Science at interface and filtering aspects because it offers advanced citation
 152 tracking features for analyzing citation patterns, detailed abstracts and faster indexing for accurate
 153 review process (Harzing and Alakangas, 2016). As a result, the Scopus database has been widely

154 utilized in previous scientometric or science mapping review articles (Antwi-Afari et al., 2023; Ye
155 et al., 2024), thus, it was selected for the present review study. A thorough search was carried out
156 using a two-part search string in the Scopus “article title/abstract/keyword” field. The first row of
157 keywords includes “digital twin” OR “digital twins” OR “virtual counterpart” OR “digital replica”
158 OR “virtual twin”, while the second row of keywords constitutes “project” OR “operation” OR
159 “maintenance”. To ensure that the articles covered the most recent years, the studied period ranges
160 from 2014 to 2024. It can also help to examine the most representative research articles for further
161 analysis.

162

163 *2.2. Stage 2 Literature selection*

164 After the literature search based on stage 1, the results must be screened to aid in identifying articles
165 within the purview of this study and credible sources that can be used for further scientometric
166 analysis. Considering the studied research domain, the publications were limited to “engineering”,
167 resulting to 3,975 publications out of 5,993 literature documents that were initially found by the
168 search query. The document type only included articles. This is because scientific articles undergo
169 rigorous peer review, and they were used to conduct the annual publication trends. Consequently,
170 book chapters, conference papers, reviews, notes, and so forth were omitted, thus obtaining 1,656
171 publications. Notably, 79 publications were excluded due to “article in press” stage, 39 articles were
172 from trade journals, and 13 book series. In total, 246 irrelevant articles were excluded because they
173 were written in languages other than “English”. After selecting all “open access” articles, 508
174 articles were obtained after the screening processes. Manual screening is imperative to narrow down
175 the application of DT in project O&M because DT has been applied in several sectors such as AEC,
176 aerospace, manufacturing and automotive, especially for flexible assembly line design or redesign.
177 Thus, articles unrelated to the DT applications in project O&M were removed. Meanwhile, other
178 unavailable articles without DOI were also deleted. Finally, 444 articles were used for scientometric
179 and critical review analysis. Table 1 illustrates the search query string and search results.

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184 **Table 1.** Search strings and results

Query String	Result
((TITLE-ABS-KEY("digital twin" OR "digital twins" OR "virtual counterpart" OR "digital replica" OR "virtual twin") AND TITLE-ABS-KEY("project" OR "operation" OR "maintenance")) AND PUBYEAR > 2013 AND (LIMIT-TO (SUBJAREA, "ENGI")) AND (EXCLUDE (DOCTYPE, "cp") OR EXCLUDE (DOCTYPE, "ch") OR EXCLUDE (DOCTYPE, "re") OR EXCLUDE (DOCTYPE, "cr") OR EXCLUDE (DOCTYPE, "bk") OR EXCLUDE (DOCTYPE, "er") OR EXCLUDE (DOCTYPE, "ed") OR EXCLUDE (DOCTYPE, "no") OR EXCLUDE (DOCTYPE, "sh")) AND (EXCLUDE (PUBSTAGE, "aip")) AND (EXCLUDE (SRCTYPE, "d") OR EXCLUDE (SRCTYPE, "k")) AND (EXCLUDE (LANGUAGE, "Chinese") OR EXCLUDE (LANGUAGE, "German") OR EXCLUDE (LANGUAGE, "Russian") OR EXCLUDE (LANGUAGE, "Korean") OR EXCLUDE (LANGUAGE, "Spanish") OR EXCLUDE (LANGUAGE, "Portuguese") OR EXCLUDE (LANGUAGE, "Slovenian") OR EXCLUDE (LANGUAGE, "Italian") OR EXCLUDE (LANGUAGE, "Polish")) AND (EXCLUDE (OA, "publisher full gold") OR EXCLUDE (OA, "repository") OR EXCLUDE (OA, "publisher hybrid gold") OR EXCLUDE (OA, "publisher free2read")))	508
Manually screening literature irrelevant to project operation and maintenance	444

185 **Note:** Scopus search was done in January 2024. The standards of irrelevant literature manual
 186 screening were described in Section 3.2.

187
 188 *2.3. Stage 3 Science mapping analysis*

189 To comprehensively understand the publications and knowledge in this field, a science mapping
 190 analysis was conducted to generate visualized scientometric network diagrams which show the
 191 graphical representation of bibliographic records. There are numerous science mapping tools,
 192 including BibExcel, CiteSpace, CoPalRed, Gephi, IN-SPIRE, VOSviewer, and many others,
 193 designed for analyzing and visualizing the bibliometric network of scientific research (Kumar and
 194 Choukimath, 2015; Wu et al., 2020). The VOSviewer tool was selected because it offers text-
 195 mining functionalities and can create co-occurrence networks from scientific literature. VOSviewer
 196 autonomously detects terms and constructs scientometric maps using web data, which provides
 197 clear graphical representation and facilitates analysis from diverse perspectives (Van Eck and
 198 Waltman, 2017). The key advantages of VOSviewer over other science mapping tools encompass
 199 its user-friendly graphical display capabilities, suitability for handling large datasets, and flexibility
 200 in accommodating diverse databases and sources in various formats (Van Eck and Waltman, 2010;
 201 van Eck and Waltman, 2017). Consequently, VOSviewer was adopted in this study to generate and

202 visualize network maps of DT applications in project O&M in order to conduct (1) keywords co-
203 occurrence analysis and (2) document analysis. Keyword co-occurrence analysis examines the
204 number of articles associated with the emerging keywords, while document analysis displays the
205 number of citations of a documents (Van Eck and Waltman, 2010). These results were further used
206 to understand and discuss the mainstream topics, research gaps and future research directions of DT
207 in project O&M.

208

209 *2.4. Stage 4 Qualitative discussion*

210 This stage will speciously analyze the contents of key selected articles related to the theme of DT
211 in project operation and maintenance, whose objective is to thoroughly assess this study's goals
212 associated with the popular study subjects in the field of DT in project O&M in the construction
213 industry. The results are obtained by VOSviewer and are based on a review of relevant publications,
214 data visualization, and temporal classification. Meanwhile, the most cited articles analysis and
215 keyword analysis can indicate the hot topics and research trends of DT application in project O&M.
216 The goal of assessing DT usage in the construction sector and establishing practical application in
217 project O&M is to highlight the difficulties and problems so that relevant researchers and
218 professionals may use them as a guide for future development directions. After the manual
219 screening, the 444 articles that passed muster will be subjected to a thorough qualitative analysis
220 that compares and discusses the annual publication trend of articles, relevant peer-reviewed journals,
221 keywords co-occurrence analysis, and document analysis. In this stage, the mainstream research
222 topics in DT in project O&M were discussed based on the keywords and identified documents in
223 the previous stages. It also articulates future research directions and research gaps that are of great
224 value to be further researched for the development of DT applications in project O&M.

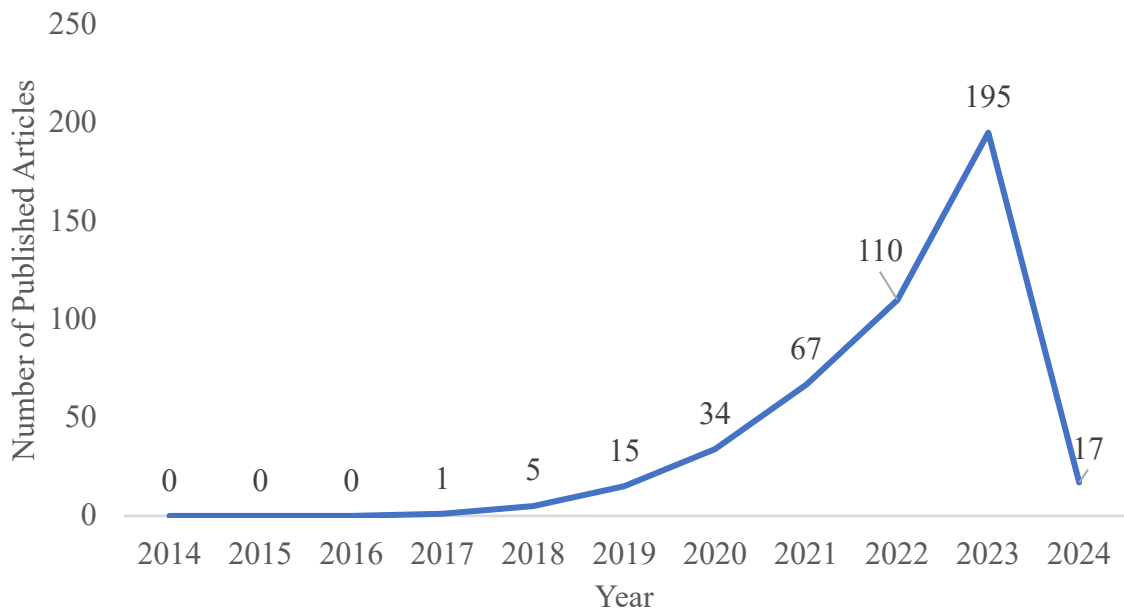
225

226 **3. Results**

227 *3.1. Annual publication trend of articles*

228 In this study, 444 journal articles were used for further analysis as distributed in Fig. 2. As such,
229 Fig. 2 shows the annual publication trend of articles related to DT applications in project O&M. As
230 shown in Fig. 2, the articles were published from 2014 to 2024. It clearly shows a significant overall
231 upward trend of number of published articles on DT applications in project O&M from 2017 to
232 2023. The increase in number of publications may be explained by the advancement of digital

233 technologies and the growing interest of practitioners and researchers in this field. For instance, the
 234 maturity of BIM applications and the development of AI technologies have promoted the demand
 235 for real-time monitoring, predictive fault diagnosis, and prevailing visualization dashboards.
 236 Likewise, the transformation of digital technologies in project management, especially the O&M
 237 stage requires more accurate and efficient digital technologies to manage facilities and their
 238 operational processes. As a result, researchers and practitioners have demonstrated the integration
 239 of BIM and DT for data modelling, assessment, and collection (Pan and Zhang, 2021; Radzi et al.,
 240 2024). It was found that the number of published articles increased in the last 4 to 5 years and
 241 peaked at 195 articles in 2023. Based on the continued growth published articles in the studied
 242 domain, it is expected that the number of publications will increase in 2024, since the data was
 243 collected as of January 2024, showing a downward trend from 2023 to 2024 with 17 articles. The
 244 future research could necessitate to confirm whether the speculation is correct.
 245



246
 247 **Fig. 2.** Annual publication trend of articles in DT in project O&M (A total of 444 articles)

248
 249 *3.2. Selection of relevant peer-reviewed journals*

250 After manual screening, 444 articles related to DT in project O&M have been selected and will be
 251 specifically analysed through the following bibliometric analysis. Table 2 shows the top-ranked 20
 252 journals according to the number of published articles based on the 444 selected articles in this

253 review study. It was found that “*Automation in Construction*” contributes to the largest number of
 254 articles at 23, occupying 5.18% of the total publications. The scope of “Automation in Construction”
 255 journal includes articles focusing on all stages of project lifecycle, with dedicated interest in digital
 256 transformation research pertaining to computer-aided decision support systems, product data
 257 interchange, facilities management, and intelligent control systems. This may be reason why
 258 researchers and practitioners who are interested in DT applications in project O&M choose to
 259 publish their articles in “*Automation in Construction*”. Six peer-reviewed journals listed in Table 2
 260 have published not less than 10 articles each, indicating that these journals also focused on research
 261 related DT in project O&M. Overall, these analyses were based on 444 articles from 174 journals,
 262 indicating that these journals can be represented as further reference values of DT research.

263

264 **Table 2.** Top 20 selected peer-reviewed journals, 2014 to 2024.

Journal name	Number of relevant articles	% of total publications
Automation in Construction	23	5.18%
Journal of Manufacturing Systems	16	3.60%
International Journal of Computer Integrated Manufacturing	13	2.93%
Robotics and Computer-Integrated Manufacturing	11	2.48%
Computers in Industry	10	2.25%
The International Journal of Advanced Manufacturing	10	2.25%
Energy	9	2.03%
Mechanical Systems and Signal Processing	8	1.80%
Sustainable Cities and Society	8	1.80%
Cirp Annals	7	1.58%
IEEE Transactions on Industrial Informatics	7	1.58%
Journal of Computing and Information Science in Engineering	7	1.58%
Reliability Engineering & System Safety	7	1.58%
Advanced Engineering Informatics	6	1.35%
Applied Energy	6	1.35%
Building and Environment	6	1.35%
Computers & Industrial Engineering	5	1.13%
Energy and Buildings	5	1.13%
Journal of Cleaner Production	5	1.13%
Journal of Computing in Civil Engineering	5	1.13%
Total	174/444	39.19%/100

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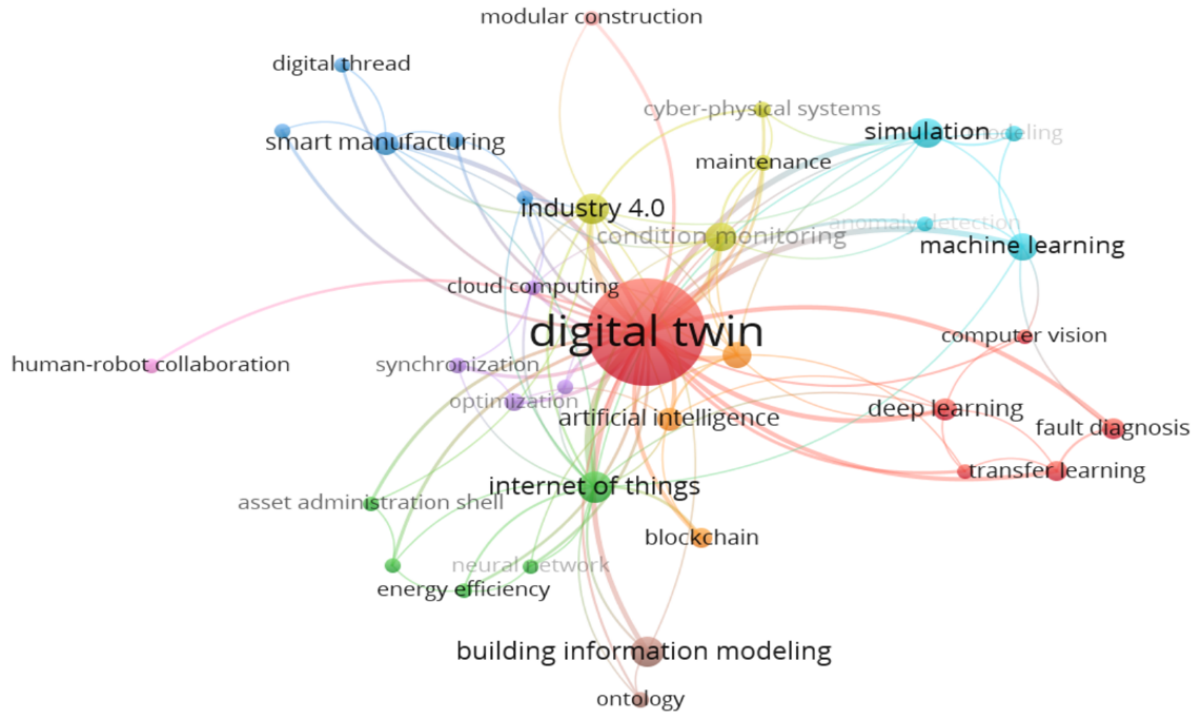
268 3.3. Keyword co-occurrence analysis

269 Keywords are representative and refined expressions of the content of a research article. Keyword
270 co-occurrence analysis allows for identifying hot topics and research areas in the knowledge
271 domain over a specific period. In VOSviewer, keywords are based on the co-occurrence analysis
272 of 444 records. The author keyword marks the keyword provided by the author. The distance
273 between the two nodes represents the strength of the relationship between the two research areas.
274 The further the distance, the weaker the relationship between each other. After merging the agreed
275 keywords using the VOSviewer thesaurus file, it was concluded that out of 1582 keywords, 35
276 items met the criteria as each word's minimum number of occurrences was set at 4. This threshold
277 was chosen after several attempts were made to obtain the best clusters for various other factors.
278 Fig. 3 shows the keyword co-occurrence network with 35 items, 105 links, and 296 total link
279 strengths.

280 The node's size indicates how frequently a keyword appeared in the data file. The top two most
281 frequently occurring keywords are "digital twin" (occurrence=320) and "internet of things"
282 (occurrence=21). These two keywords are advanced digital technologies which can be integrated
283 to enhance the digital transformation of project O&M, thus revealing why they are the most
284 frequently occurring. Additionally, certain keywords had relatively high total link strength scores,
285 which indicates a stronger connection between the keywords and the topics and themes such as
286 the top five in total link strength score, "digital twin" (total link strength=209), "internet of things"
287 (total link strength=42), "Industry 4.0" (total link strength=31), "condition monitoring" (total link
288 strength=26), and "predictive maintenance" (total link strength=22). These results indicate that DT
289 is often connected to IoT as the new paradigm of Industry 4.0 for numerous project O&M scenarios.
290 It also reveals the continued development of Industry 4.0 in a large-scale and fast-growing IoT
291 market, which utilizes IoT and DT to facilitate the realization of Industry 4.0. Nevertheless,
292 "condition monitoring" and "predictive maintenance" are the two primary critical success factors
293 for DT applications in project O&M (Jiang et al., 2021). Since Industry 4.0 promotes the agenda
294 of digital transformation, the potential of DT in project O&M has been discussed (Radanliev et al.,
295 2022). By integrating DT and other digital technologies, the condition monitoring and predictive
296 maintenance at project O&M stage could be realized, thus explaining why these keywords had
297 higher total link strength. The frequency of keyword occurrences is usually proportional to the

298 total link strength. These items also indicated that these were popular topics in the project O&M
 299 field's digital twin during the years of research.

300



301

302 **Fig. 3.** Keywords co-occurrence network of DT in project O&M

303

304 Table 3 summarises the keyword co-occurrence and each node's strength. It reveals how frequently
 305 the keywords are retrieved from the literature. It is obvious that “digital twins”, “internet of things”
 306 and “industry 4.0” appeared more frequently than others, which means these keywords are widely
 307 associated and analysed in DT in project O&M. The average publication year also reveals that the
 308 research in this area has been immensely popular in the last two years and the research intensity
 309 has been on the constant upswing stage. For example, it was only “cyber-physical systems”, “cloud
 310 computer”, and “service-oriented architecture” these two keywords occupied the year 2020, then
 311 4 keywords related to the function of DT, such as “simulation”, “modelling”, “industry 4.0”, and
 312 “maintenance” frequently emerge in 2021, more comprehensive research expansion in this area
 313 happened in 2022 (keywords=20) and 2023 (keywords=8). Keywords show the interconnection
 314 and interoperability research with other digital technologies, such as “blockchain”, “BIM”, and
 315 “IoT”, etc.

316 The links are the number of linkages with another keyword, and the total link strength can show
317 the degree of strength with a specific node. As shown in Table 3, most keywords' total link strength
318 is lower than 20. Only the more general keywords expression about the DT function in project
319 O&M, which commonly happens. VOSviewer provides a dynamic period view for keyword
320 research. Even though the keywords have been concentrated in the last four years, it is worth noting
321 that keywords such as “deep learning”, “machine learning”, “transfer learning”, and “neural
322 network” belong to the category of complex computer science and AI development. Besides,
323 “maintenance” and “cloud computing” exist relatively early and rank the highest average citations
324 at 140.8 and 84.25, respectively. It slightly implies that digitalised maintenance demand increased
325 in recent years. The technologies represented by these keywords are all linked to DT in the
326 construction industry, and the other areas of project O&M applications and development provide
327 the basic equipment conditions and data model basis.

328 Regarding the average normalised citations, “cloud computing” still ranks the highest at 2.68 in
329 Table 3. This means that these keywords are cited the most on average. It shows that the average
330 citations are disproportionate to the average normalised citations. For example, the average
331 normalised citation of “smart manufacturing” is the highest, and the highest keyword based on
332 average citations is “maintenance”. It indicates that computer science and complex digital
333 technologies have changed traditional operation methods and have become more intelligent for the
334 construction and manufacturing sectors. Using advanced robotics and model work techniques will
335 enhance the need to adapt human-machine interaction strategies, interconnection of multi-
336 technologies, and workflows.

337

338 **Table 3.** List of keywords of co-occurrence analysis

Keywords	Occurrences	Links	Average publication year	Average citations	Average normalized citations	Total link strength
Digital twin	320	34	2022	20.25	1.06	209
Internet of things	21	18	2022	27.90	1.30	42
Industry 4.0	20	12	2021	37.30	1.35	31
Condition monitoring	17	10	2022	42.35	0.81	26
Predictive maintenance	14	11	2022	35.43	0.98	22
Simulation	18	8	2021	15.94	0.37	21
Machine learning	16	7	2022	18.75	0.89	19

Smart manufacturing	11	7	2022	44.27	2.20	16
Deep learning	10	7	2023	18.80	0.88	15
Artificial intelligence	11	7	2022	23.73	1.29	14
Building information modeling (BIM)	20	4	2022	32.20	1.45	13
Cloud computing	4	7	2020	84.25	2.68	12
Fault diagnosis	9	3	2023	4.22	1.22	11
Transfer learning	8	4	2023	3.38	0.49	11
Blockchain	8	3	2022	32.00	2.02	10
Sensors	4	7	2022	11.50	0.36	10
Optimization	6	4	2022	6.17	0.52	8
Synchronization	5	4	2022	22.40	1.14	8
Cyber-physical systems	5	4	2020	67.60	1.23	7
Maintenance	5	4	2021	140.80	2.44	7
Energy efficiency	4	4	2022	19.00	1.37	7
Fatigue	4	4	2022	9.75	0.77	7
Remaining useful life	4	4	2023	5.75	0.67	7
Service-oriented architecture	4	4	2020	48.75	0.94	7
Anomaly detection	4	3	2022	29.75	1.54	6
Asset administration shell	4	3	2022	26.25	0.64	6
Neural network	4	4	2023	2.25	0.78	6
Virtual reality	5	3	2022	3.00	0.37	5
Modular construction	4	2	2022	12.50	1.70	5
Reinforcement learning	4	2	2023	2.75	0.95	5
Modeling	5	3	2021	17.20	0.44	4
Ontology	5	3	2023	9.20	0.97	4
Computer vision	4	3	2023	6.50	0.89	4
Digital thread	4	2	2022	12.75	1.75	4
Human-robot collaboration	4	1	2022	14.00	1.11	3

339

340 3.4 Document analysis

341 Setting the minimum citation number of a document to 50, 34 meets the threshold out of 432
342 documents. The detailed top 15 most normalised citation articles have been listed in Table 4, and it
343 was found that Feng et al. (2023) study got the most normalised citations at 27.15 but relatively

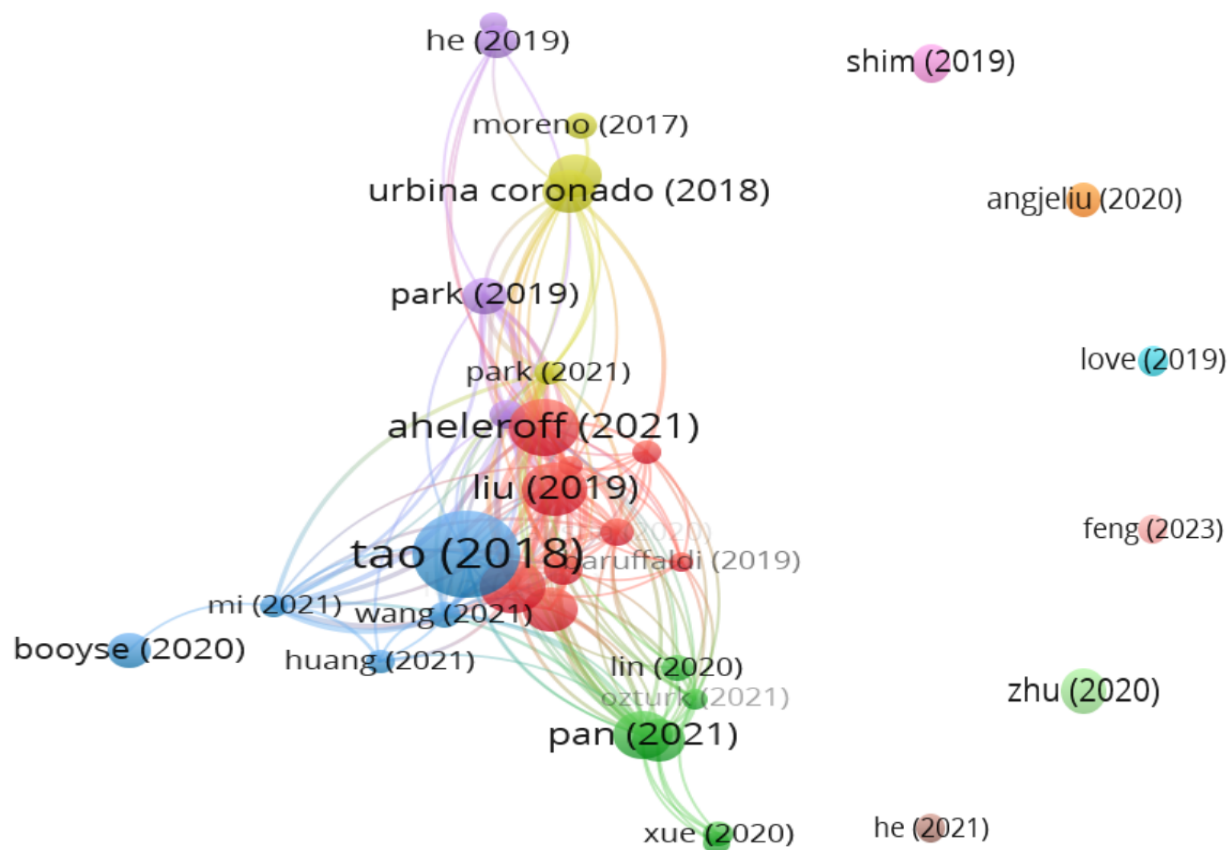
344 middle-level citation numbers at 82. This article was published in 2023, the newest among other
345 articles, so that it might be the reason for the lower citation number. It indicates that DT significantly
346 affects asset fault diagnosis or condition monitoring through intellectual property. Some articles
347 integrated DT with BIM and other technologies like blockchain and human-robotic interaction to
348 increase the efficiency of the construction and manufacturing industries. From Table 4, the articles
349 of Pan and Zhang (2021), Lee et al. (2021), and Aheleroff et al. (2021) are all related to project
350 management and service system design, which expresses the DT application research is more
351 focused on the managerial process and procedures, especially in the year of 2021.

352 Subsequently, the articles in Table 4 indicated that the DT research has integrated with the existing
353 technologies implemented in the construction industry, such as BIM, IoT, and blockchain
354 technology. To better develop DT in project O&M in this industry, He et al. (2021), Lee et al.
355 (2021), Zhu et al. (2020), and Wang et al. (2021, 2022) developed DT integration with other
356 technologies to enhance the historical data support and existing technologies application efficiency.

357 Likewise, smart city design and management are driven by the DT and incumbent digital
358 technologies to control traffic and monitor city conditions, which can be shown by the articles of
359 Xia et al. (2022) and Zhu et al. (2020). Besides that, the articles about DT function in project O&M,
360 such as asset management, facilities health condition prediction, and fault diagnosis, become more
361 crucial and mature, even connecting with specific case studies. This can be seen in the articles of
362 Luo et al. (2020), Tao et al. (2018), Stark et al. (2019), and Booyse et al. (2020).

363 Fig. 4 illustrates the document network analysis. From this figure, the front and size of the nodes
364 stand for the citation number of each article. It was found that Tao et al., 2018 got the most citations
365 at 489 with a lower level of normalised citations at 2.86. This article was published in 2018, which
366 is a relatively earlier publication. As such, this article received much attention in the expertise field.

367 As shown in Fig. 4, many single nodes represent many articles with fewer citations and no links to
368 other articles, which may also be a sign that the research topic of these standalone articles still needs
369 to receive more attention than other articles.



370
371 **Fig. 4.** Document network analysis of DT in project O&M
372

373 **Table 4.** Top 15 cited articles among 444 articles, 2014-2024

Articles	Titles	Citations	Normalized Citations
Feng et al. (2023)	Digital Twin-Driven Intelligent Assessment of Gear Surface Degradation	82	27.15
Li et al. (2022)	Digital Twin in Smart Manufacturing	164	11.72
Aheleroff et al. (2021)	Digital Twin as a Service (DTaaS) in Industry 4.0: An Architecture Reference Model	257	8.44
Pan and Zhang (2021)	A BIM-Data Mining Integrated Digital Twin Framework for Advanced Project Management	193	6.33
Lee et al. (2021)	Integrated Digital Twin and Blockchain Framework to Support Accountable Information Sharing in Construction Projects	166	5.45
Luo et al. (2020)	A Hybrid Predictive Maintenance Approach for CNC Machine Tool Driven by Digital Twin	227	4.42
Xia et al. (2022)	Study on City Digital Twin Technologies for Sustainable Smart City Design: A Review and Bibliometric Analysis of geographic information system and building information modeling integration	60	4.29

Zhu et al. (2019)	Parallel Transportation Systems: Toward IoT-enabled Smart Urban Traffic Control and Management	166	3.23
Liu at al. (2019)	Digital Twin-driven Rapid Individualised Designing of Automated Flow-Shop Manufacturing System	227	2.88
Tao et al. (2018)	Digital Twin Driven Prognostics and Health Management for Complex Equipment	489	2.86
He et al. (2021)	BIM-enabled Computerized Design and Digital Fabrication of Industrialized Buildings: A Case Study	84	2.76
Wang et al. (2021)	Digital Twin for Human-Robot Interactive Welding and Welder Behavior Analysis	80	2.63
Booyse et al. (2020)	Deep Digital Twins for Detection, Diagnostics and Prognostics	124	2.41
Stark et al. (2019)	Development and Operation of Digital Twins for Technical Systems and Services	177	2.24
Park et al. (2021)	The Architectural Framework of a Cyber-Physical Logistics System for Digital-Twin-Based Supply Chain Control	67	2.20

374

375 **4. Discussion**

376 After reporting the results of this review study, this section will mainly focus on the main research
377 topics, research gaps, and future research studies within the theme of DT in project O&M. Although
378 previous studies on DT were primarily conducted in other disciplines such as automobile,
379 manufacturing, the current review paper focuses on the construction sector, which is the research
380 hotspot topics in recent years and caused significant interests of scholars and industrial participants.
381 The construction sector is typically project-based, its O&M involves multiple stakeholders and
382 complex project task requirements. Nevertheless, we cannot discount the value that DT will bring
383 to the construction sector and the development of project management, which also affects the
384 reference value of this review paper.

385

386 *4.1. Summary of the main research topic of DT in project O&M*

387 The main keywords of the given topic have been listed in Table 3. There are inner links between
388 keywords such as DT, deep learning, computer version, and fault diagnosis. In most occurrences,
389 more keywords are related to other emerging technologies, such as blockchain technology, AI, BIM,
390 virtual reality (VR), etc. These are the basis of industry intelligent transformation, enabling
391 changing the O&M procedure and processes while operating real production activities. Section 4.4

392 has illustrated the detailed keyword clusters, and the mainstream themes and topics can be
393 summarized below.

394

395 *4.1.1. DT-based artificial intelligence (AI) technology for project O&M*

396 Emerging technologies such as AI and its function enrich DT application efficiency and advantages,
397 which attract significant attention. As shown in Table 3, AI, deep learning, transfer learning, and
398 ML are all ranked at a relatively higher occurrence level. DT enables the real-time virtual and real-
399 world connection with the fault diagnosis and prediction functions to optimize the project O&M
400 processes and outcomes. Based on data science and AI technology development, complex and large
401 numbers of engineering machines and equipment data can be analyzed in different AI models. For
402 example, deep learning, as the new generation of AI technology shows obvious advantages in
403 feature extraction, intelligence level, and knowledge-learning aspects (D'Urso et al., 2024). Many
404 experts are exploring integrated DT with their specific AI models to monitor machine health
405 conditions and predict their longevity through historical data dynamic comparison (Zhang et al.,
406 2023c), then facilitating the reliability of managers' O&M decision-making. DT provides the
407 simulation environment to train the AI algorithms and models, which is similar to people learning
408 knowledge to solve O&M problems (Bordegoni and Ferrise, 2023). For example, to increase the
409 DT data efficiency and accuracy, control the data collection costs, solve data latency problems (Gao
410 et al., 2023), and enlarge the DT application scale in the real industrial production process, Xia et
411 al. (2023) developed DT integrated with transfer learning and cloud-based model to diagnosis the
412 faults of DC/DC converter to support power supply systems maintenance strategy setting in time.
413 On the other hand, the simulation model and analysis of ML require plenty of time to execute them,
414 which limits the models' application, but DT's virtual world duplicates the physical world becomes
415 a perfect place to modifier and develop AI models application, then improve the production and
416 operation efficiency (Jain and Narayanan, 2023). These two emerging technologies integration
417 solve application and research challenges from both sides and generate reciprocal effects.

418 While project O&M is the longest and most complex project stage, it usually involves multiple
419 machinery equipment, facilities, people engagement, and daily building space systems management
420 to ensure the function of the building executes efficiently and fulfills the demands of project
421 customers and users (Zhao et al., 2022). Table 3 also incorporates the keywords like “cloud
422 computing”, “remaining useful life” and “service-oriented architecture”. AI and DT technologies

423 application sufficiently renew project O&M methods and generation, facility and project managers
424 are assisted by these emerging technologies to design equipment management and operation plans
425 and support intelligent, scientific, feasible decision-making on maintenance strategies (Zhang et al.,
426 2023a). The powerful arithmetic of AI can replace manual information search and analysis. For
427 instance, when emergency circumstances happen in the building, DT models can align with BIM
428 (Pan and Zhang, 2021), information communication technology (ICT), and mixed reality (MR) (Wu
429 et al., 2022) etc., to aid real-time workforce safety identification, facilities health monitoring, and
430 fault diagnosis in a real-time.

431

432 *4.1.2. DT-enabled smart city and sustainability*

433 Since emerging technologies have brought changes to the traditional construction methods and
434 concept of urban architecture. From a macro perspective, smart city O&M also plays a crucial role
435 in space intelligence. Given the advantages of DT technology for real-time operational and fault
436 prediction capabilities, numerous areas are investigating in-depth applications for smart
437 manufacturing, smart energy, and smart homes. The smart city relies on IoT, GIS, BIM, AI, and
438 other technologies in the industry 4.0, and 5.0 environment to reduce operational human
439 involvement through an intelligent method to realize the sustainable urban O&M, it is a symbol of
440 the urbanization process which increases the security, cost-saving, and intrinsic connections of the
441 city (Silva et al., 2018). Bujari et al. (2021) designed a distributed geographic system for different
442 cities' heterogeneous data absorption and analysis to consider potential interruptions of daily city
443 operation stakeholders through DT, cyber-security systems, and big data.

444 Another research theme is traffic network optimization, as the climate change and traffic flow issues,
445 DT technologies are associated with GIS and other geographic technologies to realize real-time
446 traffic network monitoring and modification. AI and IoT-supported smart traffic systems were
447 developed by Zhu et al. (2019) through a large number of iterative data simulations over a long
448 period are used to predict the operational outcomes that may come out of specific cases, and then
449 to plan, design, and operate and maintain the smart urban traffic system control. City transport
450 infrastructure maintenance, such as bridges, roads, and railways, all can be operated and maintained
451 through DT-enabled data-driven methods (Wang et al., 2023). In addition, the keywords such as
452 “anomaly detection” (Lu et al., 2020b), “sensor”, “synchronization”, etc. emphasize the real-time

453 data update and facilities fault diagnosis function of DT, so there are some researchers are
454 exploring DT applications for city heritages maintenance.

455 Sustainability indicates the reducing of product lifecycle emissions, pollution, and consumption,
456 and the improvement of the accompanying environmental, economic, and social benefits (Zhang
457 et al., 2021). DT utilized for energy performance monitoring and prediction maintenance has been
458 broadly discussed and researched as the key part of energy system digitalization and optimization.
459 Zhang et al. (2023b) renewed the heating, ventilation, and air systems of heritage buildings
460 through sensor configuration to enhance energy consumption and air quality management.
461 Numerous articles discussed how industry 4.0 has impacted city design and construction due to
462 issues with resource utilization, increased energy efficiency and hazardous waste disposal, and
463 living aspects, then facilitating city sustainability O&M (Safiullin et al., 2019).

464

465 *4.1.3. DT applications for project asset management*

466 DT application in project asset management has been broadly explored in the manufacturing,
467 energy, and construction sectors (Edwards et al., 2023). Table 3 presents keywords like “asset
468 administration shell” and “human-robotic collaboration” articulating the essential function of DT
469 in asset management. As civil engineering inevitably requires equipment/plant, how to efficiently
470 utilize and maintain this large equipment is also a germane topic. DT integrated with the IT systems
471 and BIM models to upload the operation costs details, assets health commission brochure
472 generation for operators, and then construct a user-centred dashboard for various stakeholders to
473 master asset status from the whole lifecycle (Keskin et al., 2022). Besides that, the operational
474 workflow design in the DT environment not only relies on sensors’ data transmission but also on
475 assets' attributional and historical data consisting of the asset administration shell for the asset
476 operators to understand the data-driven workflows and demands (Grüner et al., 2023). DT
477 applications for asset management still need to consider the operational cost and capital cost
478 benefits and how digital technologies transform traditional commercial activities to achieve more
479 business value (Love and Matthews, 2019).

480 Although human-robot collaboration (HRC) has been discussed over recent years, there are still
481 more studies that need to be conducted by combining the recognition and physical levels of humans
482 and robot collaboration to realize common goals (Sun et al., 2022). Current research focused on
483 ML technology utilization in the process of HRC. Semeraro et al. (2023) stated the categories of

484 collaboration tasks and claimed the importance of the usage of time-dependency ML. Wang et al.
485 (2024) developed a framework to train DT data based on a neural network, then test it in the physical
486 system to improve the feasibility of human-robotic collaboration safety. Choi et al. (2022)
487 mentioned an original creation of an integrated system based on mixed reality (MR), deep learning,
488 DT, 3D point cloud data, and HRC to real-time secure human safety while working which can be
489 implied in the daily project O&M scenes. Safety management should be updated within the HRC
490 working framework because the safety standards have changed. The HRC's relevant safety
491 standards such as DIN ISO 10218-1&2 or ISO TS 15066, which can implicate different sorts of
492 standards can complicate the operational procedure and require more measurement efforts. DT as
493 the duplication version of the physical world provides a perfect platform for safe robotic algorithm
494 development (Baratta et al., 2023).

495

496 *4.1.4. Blockchain integrated DT for project O&M*

497 Blockchain can be regarded as the distributed ledger technology and asymmetric encryption
498 technology, which ensures that decentralized transaction data can be recorded and shared traceably,
499 immutably, and transparently (Adu-Amankwa et al., 2023). Each participant can be a node, which
500 is operated by computer servers. Then, the smart contract is a digital type of contract codification
501 and execution, once the conditions of the smart contract are met, the payment mechanism is
502 automatically triggered. Blockchain has been broadly explored in Finance, Retail, Public
503 institutions, and construction sectors. Due to the fragmented, complex, and uncertain features of
504 construction projects, blockchain-enabled information sharing and decentralized organizational
505 structure can help tame complexity (Papadonikolaki and Jaskula, 2023) and ensure computer
506 algorithm-based governance. DT transports the building information in real-time to the blockchain,
507 then all the transaction data is traceable and immutable recorded with a timestamp, which facilitates
508 various project stakeholders' responsibility clarification and understanding (Lee et al., 2021). A
509 study by Lee et al. (2021) also ranked at a high level in Table 4. Blockchain sometimes is mainly
510 used to guarantee the trustworthiness of DT data, for its historical data will be overwritten in the
511 dynamic asset O&M process (Tavakoli et al., 2023).

512 In addition, blockchain-enabled DT technology assists cross-disciplinary stakeholders'
513 collaboration and breaks information island in the process of construction, resource leveling,
514 information tracking, and task execution guidance and alarm can all be utilized by DT in a timely

515 order (Jiang et al., 2023). Contract management, project quality management, and even the whole
516 lifecycle management can be shaped by blockchain-enabled DT technology. In a word, blockchain
517 enables collaboration within the heterogeneous social, network, and physical space resources (Li et
518 al., 2023).

519 However, blockchain and DT mobilized the transaction methods and information-sharing traditions,
520 but as the disputed technologies, the project-based organization structure and governance
521 approaches influence stakeholders' relationship establishment also attract many discussions. Qian
522 and Papadonikolaki (2021) concluded that blockchain can promote trust relations from many
523 dimensions and shift relation-based trust to cognitive and system-based trust in the supply chain
524 management field. Blockchain and DT utilization as a decision-making support tool in the complex
525 logistic and inventory management process exerted an efficient function as well (Pan et al., 2021)
526 (Gai et al., 2022). Subsequently, stakeholders' acceptance, operation capability, and technological
527 understanding still hinder the real application in the construction sector.

528

529 *4.1.5. DT for advanced project management*

530 The digitalization degree of construction project management is still under a transition process.
531 Digital technologies such as IoT, BIM, VR, MR, ICT, and others all contribute to the transition
532 progress while DT exceedingly promotes the development of advanced project management with
533 all these digital technologies. It enables the project O&M process to automate and eliminate the
534 procedures that involve human intervention to construct event diary logs and evaluate procedures'
535 performance (Pan and Zhang, 2021). There are some keywords and articles primarily pertinent to
536 the function of DT in Table 3 and Table 4, such as "fault diagnosis", "predictive maintenance",
537 "condition monitoring", "Synchronization", "Service-oriented architecture", "fatigue",
538 "optimization", and "anomaly detection", which indicates how DT contributes to project O&M,
539 then, further to advanced project management. Feng et al. (2023) designed a DT model to assess
540 the gear surface degradation situation, whilst Booyse et al. (2020) studied the trace degradation of
541 the asset lifecycle situation through response to the asset data simulation model. A study by Dreyer
542 et al. (2021) implemented DT to examine the energy pipelines' fatigue and health situation, instead
543 of periodic stopping of equipment for piping equipment assessment.

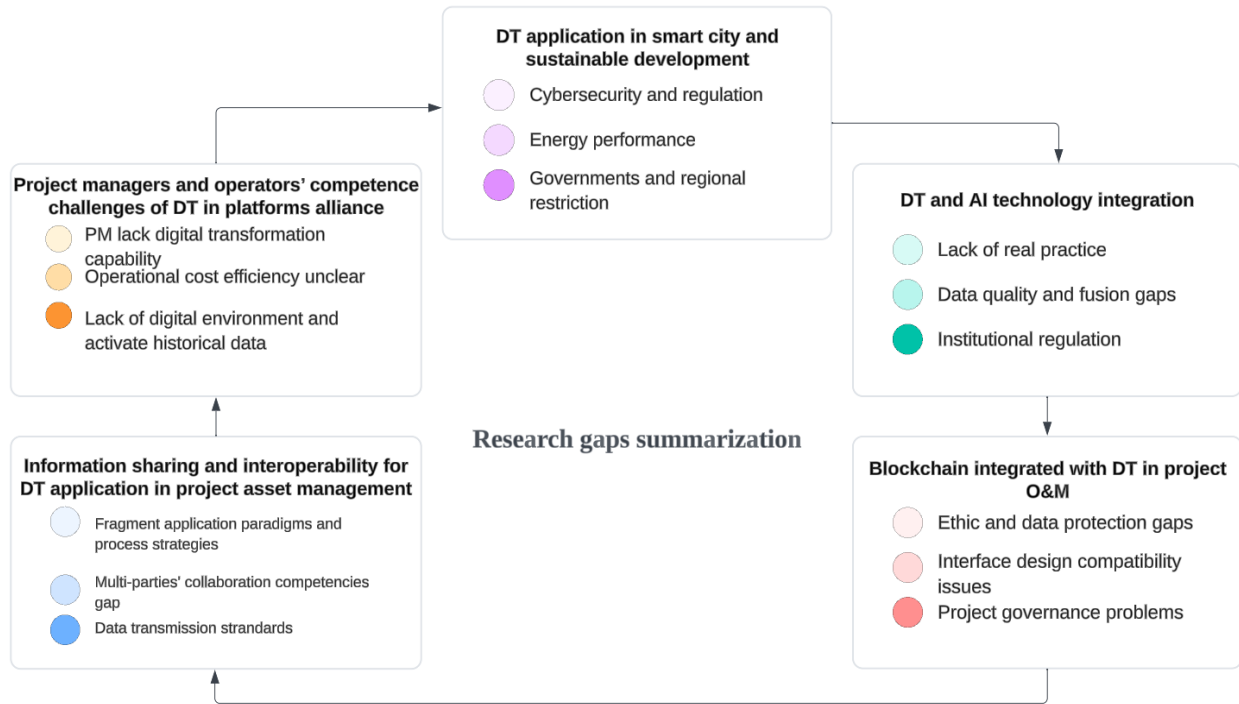
544 Project complexity and uncertainty always challenge the performance and success of projects,
545 O&M is the longest stage, and the complex degree is relatively high, many articles also mentioned

546 modular construction integrated with VR (Wu et al., 2022) and DT. DT can be used to monitor on-
547 site assembly situations and the modular installment process. DT and VR as the immersive training
548 and simulation platform for the robot arrangement in the modular construction will mitigate the
549 construction on-site complexity degree and increase the safety measurements (Zhu et al., 2022).
550 In the case of decision-making bias and optimization bias of project operators and managers, DT-
551 driven decision-making is based on event simulation and precise data calculations, which enhance
552 the feasibility and accuracy of decisions. For example, Fang et al. (2024) developed a highway
553 infrastructure rebuild, recovery, and operation decision support visualized platform and framework
554 by DT for the project and government decision-makers. Decision-making is one of the key elements
555 of project governance, which will severely affect the project's performance and success, as well as
556 the satisfaction degree of end users and project clients (Müller-Zhang et al., 2023). Risk
557 identification and mitigation strategy suggestions provided by DT real-time structure health
558 monitoring and predictive maintenance then assisted the service-oriented structure construction
559 (Shi et al., 2023). Project O&M processes are generally becoming a service provision process. DT
560 implementation conspicuously improves the efficiency and productivity of project O&M, then
561 contributes to the advanced project management knowledge fields and empirical project practice.

562

563 *4.2. Research gaps of DT application in project O&M*

564 The following subsections describe in detail the six main research gaps of DT application in project
565 O&M based on the identified mainstream research topics. The contents of each are presented in
566 Figure 5, which briefly summarizes the limitations or challenges of DT application.



567

568 **Fig. 5.** Research gaps of DT application in project O&M

569

570 *4.2.1. Scenarios of DT and AI technology integration in project O&M*

571 DT and AI technology have been utilized in many areas, such as smart manufacturing, smart cities,
 572 and aerospace fields. There are three main stages of AI+DT implementation in project O&M. The
 573 first stage is monitoring and observation, through the DT platform to realize the whole project
 574 lifecycle and asset health condition monitoring in a real-time, complex project O&M scenarios
 575 problems can be found and fixed with the power of arithmetic and knowledge of AI technology.
 576 The second stage is prediction and forecast, after detailed data collection, the simulation models
 577 of project participants and O&M equipment can be built to get the simulation results for different
 578 project tasks and events. The third stage is evaluation and decision optimization. The multiple
 579 simulation possibilities calculation and comparison assisted by deep learning etc. technologies to
 580 design optimized plans and strategies (Lv and Xie, 2022). However, the intelligent project on-site
 581 construction environment and procedure planning require a large amount of data on building,
 582 workflow, and operators' collaboration. Many AI+DT applications in project O&M primarily
 583 proposed concept models without real practice and testing in industrial production scenarios
 584 (D'Urso et al., 2024). From the managerial angle, the analysis capability of the AI systems can

585 only execute the designed tasks, but lack of lesson learned reflection, performance enhancement
586 depends on the historical data accuracy and completeness. The general project management
587 regulation for AI and DT applications needs to be refined (Müller et al., 2024).

588

589 *4.2.2. Challenges for DT application in smart city and sustainable development*

590 By 2050, the urban population is expected to increase by 2.5 billion, according to the United Nations,
591 it can be inferred that the significance of smart cities and the associated intelligent solutions will
592 rise significantly, even though changes in how cities function and live are closely related to the
593 region's politics, economy, traditions, and culture (Popović and Rajović, 2021). Industry 5.0 will
594 promote the intelligent degree of the population's urban life, but the traditional architectural
595 structure limits the implementation of emerging technologies, furthermore, it will affect the
596 realization and sustainable development of the smart city. Smart cities necessitate a protracted
597 process for changing how current urban buildings are maintained and operated, but the
598 contemporary architectural designs' capacity for sustainability, energy conservation, extreme
599 weather, and geohazard prevention and control will undoubtedly grow (Goyal et al., 2020). Digital
600 transformation of incumbent buildings will be certainly constrained by governments of different
601 countries, technical feasibility, social collaboration, and economic benefit (Tomičić Pupek et al.,
602 2019). Hence, it is necessary to consider the detailed smart city initiation and supporting O&M
603 measures as much as possible in the construction design phase, and the data-oriented lifestyle must
604 be suitable for future technology upgrades. Meanwhile, it is imperative for the designer and
605 construction workers to consider the overlapping and duplicating parts of technical functions to
606 avoid high input costs and low economic efficiency.

607 Smart city statistics accumulated for DT model construction are essential, but cross-region city
608 data collection and interconnection are under development due to data regulation and different
609 laws. As the DT system involves various specific city operating data, while urban safety and
610 administration systems are distinct, it is harder to support other institutions' usage based on data
611 demands, therefore, cross-institution and cross-region DT data framework is required for city
612 information resource integration and fully visualized city operation process planning and
613 management. However, for highly connected smart cities, high-tech reliance increases the concern
614 about cybersecurity. Urban infrastructure O&M with DT applications face many challenges in the
615 field of cybersecurity and regulation, such as citizens' personal data protection, regulations, policy

616 clarification standards based on different categories of human behavior, and intelligent city models'
617 access authority. It is also crucial for policymakers to consider the regulation tools and policies to
618 promote city stakeholders' participation in smart city development and the daily O&M processes
619 (Almeida, 2023).

620

621 *4.2.3. Issues of information sharing and interoperability for DT application in project asset*
622 *management*

623 Towards the aim of realizing more efficient and comprehensive DT applications in project O&M,
624 information and data are the primary elements, especially for empowering the interface between
625 DT and multiple engineering technologies. As mentioned in previous sections, the data latency,
626 accuracy, and source of historical data are all breakthroughs for DT applications in practice. The
627 sensors' data transmission for DT system construction, the various sources of data integration with
628 multiple technologies, the large amount of data empowerment, and interoperability problems would
629 impact the coordination effects of these digital technologies. Collection of cyber-physical systems,
630 VR/MR, Digital modelling, and IoT make it possible to upload data and knowledge for diagnosis
631 tasks based on mega-data analysis. The whole process around data which indicates assets data
632 collection, materials, and systems' compatibility is the core resource for the quality of DT function
633 application. Therefore, the effective integration of the DT data model with the sensor data and its
634 compatibility with the machinery equipment command reception system, the quality of the data,
635 the data standards, and the restrictions on the level of authority between systems are all key factors
636 in the O&M stage of the project.

637 Although previous studies have demonstrated HRC, it is the future of industrial production works,
638 and any emerging technology must consider how to coordinate with humans and integrate with
639 existing working processes and systems. One of the HRC difficulties is the dynamic nature of
640 human behavior and the complexity of cognition is difficult to match perfectly with the need for
641 data-driven adaptation (Malik and Brem, 2021). So, AI technology simulates the human behavior
642 trajectory occupied one of the key commission and testing solutions (Wang et al., 2024), but it is
643 imperative to create a framework or lead concept for industries to describe the technologies,
644 structures, information flow, and processes that redefined operation methods. The risk assessment
645 standard should be suitable for the HRC operational environment based on the new HRC safety
646 standards. Wilhelm et al. (2021) mentioned that DT enables machines or robotics to concentrate

647 on the safety of operators through integration with a VR environment in collaboration with a
648 robotic arm to detect safety hazards and provide risk mitigation action immediately. The whole
649 process is a two-way interaction and information exchange between the operator and the robot.
650 Complex industrial environments and multiple operators' collaboration with the robotics and
651 gesture evaluation is still challenging (Wang et al., 2024). DT is prone to realize interdisciplinary
652 knowledge integration, however, the uniform convergence of application paradigms and process
653 strategies is still fragmented (Fan et al., 2021).

654

655 *4.2.4. Problems of blockchain integrated with DT in project O&M*

656 To solve the data overwritten and reliability problem of DT, blockchain technology has been
657 broadly discussed and trying to fix the data-related problem. Tavakoli et al. (2023) developed a
658 DT data resource model based on the Remix Ethereum platform and Sepolia test network which
659 enables asset model maintenance and real-time data acquisition in the dynamic O&M process.
660 Blockchain reduces fraud activities and opportunism risks by accurately tracking asset information
661 and data, which increases asset evaluation accuracy. However, the main problem is to test
662 blockchain technology in real practice, promote suppliers' and stakeholders' understanding of this
663 technology, and ensure strong data-synchronized with tele-infrastructure (Li et al., 2023).

664 In addition, because of the privacy nature of smart contracts and transaction data, cybersecurity
665 and data protection problems should be given more attention when sharing information on
666 blockchain platforms. Since most DT applications involve physical, optimized, simulated
667 knowledge and technological models, intellectual property rights should be protected.
668 Consequently, the regulation of blockchain technology and DT applications for commercial
669 purposes should focus on the ethical consent and data monopoly problems associated with their
670 development and implementation (Jiang et al., 2023). Clear regulation and governance rules for
671 blockchain technology applications should be distinguished from other digital technologies, due
672 to its decentralized transaction and algorithm-based trust mechanisms in the construction project
673 supply chain management (Gai et al., 2022).

674

675 *4.2.5. Project managers and operators' competence challenges of DT platforms alliance*

676 The level of professional competence of system operators and the complexity of contextual
677 information also pose challenges for multi-party collaboration. The characteristics of construction

678 project stakeholders, cross-organizational collaboration, and temporary natures sometimes limit
679 the digital transition acceptance. As such, complex project O&M tools like DT and AI need to
680 develop a user-friendly dashboard in industrial practice. In addition, ease and knowledge
681 understanding degree affect digital acceptance when using information and communication tools
682 is the most significant factor. Project decisions would be directly impacted by changes in tool use.
683 For example, the collaboration between various enterprises, their data and information analysis,
684 processing capabilities, and compatibility differ. During the process of bidding, business plan
685 evaluation, etc., herd decisions may occur. These decisions will impact various organizations and
686 pose the greatest threat to rational decision-making (Shi et al., 2023). The decision maker's
687 attention poses the biggest challenge to logical decision-making.

688 Researchers have discussed the success elements of how DT facilitates production and
689 sustainability, but the critical factors of DT adaptation in the organization or operational procedure
690 need further arguments (Deepu and Ravi, 2021). The key element of technology adoption is
691 typically related to people who have the digital awareness and capability to conduct holistic
692 implications and regular checks of the operational conditions. It requires high professional
693 expertise for project managers to be equipped with the knowledge of organizational processes and
694 O&M strategies, and the ability to provide a pivot and leadership role for digital transformation
695 and DT adoption. This provides challenges for project management practitioners and advanced
696 project management development. Especially, more and more emerging technologies have been
697 implemented and explored in the construction industry, and this industry particularly depends on
698 accurate events, risk, and cost budget control in project planning phase (Regona et al., 2022). High-
699 end technologies like AI and DT utilized in the scheduling or O&M phases can reduce the time
700 spent on repetitive tasks. Therefore, the project managers and experts establish a holistic digital
701 environment and history data activation is still a daunting task.

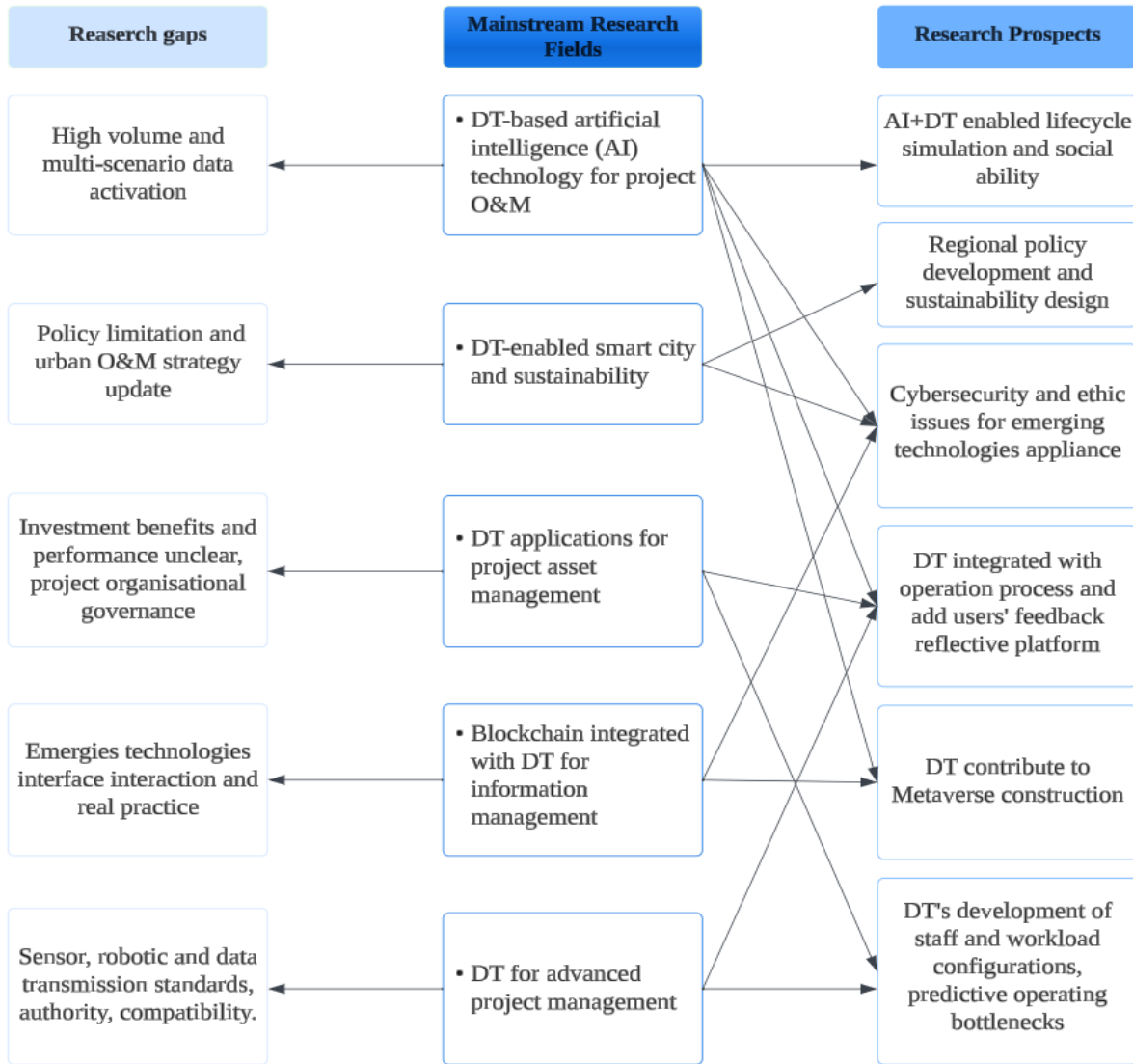
702 Finally, the efficiency of project O&M costs and capital costs for introducing DT technologies will
703 increase, but the investment performance and benefits are still unclear. These limitations hinder
704 the implementation progress of DT, the whole planning, management, and procurement process
705 of digital systems are meant to conform with the project's financial capabilities. The
706 commercialization of DT is concerned with the market and industrial participants.

707

708 *4.3. Future research studies of DT applications in project O&M*

709 After the critical literature review, science mapping analysis, and qualitative discussion, the main
 710 research themes for DT in the project O&M have been identified. Notably, the identified
 711 mainstream research topics, research gaps, and future research directions are interlinked and
 712 progressive which can be shown in Figure 6. This section is based on the former two subsections'
 713 discussion to obtain the future domain research trends.

714



715

716 **Fig. 6.** Research framework showing the links between mainstream research fields, research
 717 gaps, and research prospects

718

- 719 1. **Smart City Scenario.** Smart city design, construction, and change cannot rely solely on
720 industry-specific technology adoption and application. Regional policies can directly
721 influence the state of smart city development, and the realization of smart cities relies not
722 only on technology integration and consideration at the design stage but also on regional
723 policy innovation (Hervás-Oliver, 2021), industry-related stakeholders, and expert
724 participation. Smart city scenarios design and consideration can expand to the global
725 perspectives, and reflect multi-attitudes of citizens, policymakers, and industries. Energy
726 performance, intelligent agriculture, sustainability, etc. city development agenda will be
727 considered first in DT simulation systems.
- 728 2. **Asset Data and information aggregation.** Project managers should improve their
729 competence to comprehend and apply digital changes to organizational equipment and
730 operating procedures. The operation of deploying predictive asset management models and
731 the use of tools should consider the complexity of human involvement and behavior and
732 mitigate the detrimental effect on rational decision-making in the name of efficiency. The
733 integration of various technologies and apparatus throughout this time will unavoidably
734 raise several issues that call for further debate, including system compatibility, the adaption
735 of different machine interfaces, data fusion capabilities, and the best alignment of
736 operational procedures. The investment efficiency on DT kinds of technologies should
737 align with other digital technologies expenditure and their benefits for facilities
738 management.
- 739 3. **AI and DT integration.** It is recommended to integrate data and systems of the whole
740 construction ecosystem to establish a holistic project lifecycle and supply chain digital
741 dynamic simulation. It is imperative to break the information island limitation and
742 empower structural data. ML and deep learning technologies will be integrated with DT
743 for human gestures and behavior prediction when collaborating with robotics (Wang et al.,
744 2024). AI social ability development can predict human emotions, behavior, and thoughts,
745 which will be utilized for risk, change, cost, quality, and contract management fields
746 (Müller et al., 2024).
- 747 4. **Privacy and security.** Cybersecurity and regulations will be more robust and assisted by
748 DT as a reference when designing, commissioning, executing, and terminating policies or

749 industrial regulations. Ethical consideration and personal data protection measures will
750 also be integrated into DT and other emerging technologies development.

751 5. **O&M Scenario Consideration.** DT encourages the revitalization of information
752 management, but more research is needed to understand how data information can be used
753 to enhance physical workflow and staffing, identify operational and maintenance
754 bottlenecks, plan staff workloads, and apply employee data and job matching in smart ways.

755 6. **Decision-making Support.** In the future, the decision-making optimization will be
756 enriched because of the incorporation of users' feedback on DT and the feedback platform
757 in DT to assure resource delivery and facilitate the interaction between the virtual world,
758 physical world, and human's social world (Rožanec et al., 2022). DT and blockchain
759 technologies will contribute to the metaverse construction.

760
761
762

763 5. Conclusions

764 This review paper applied a scientometric analysis method which consists of a literature search,
765 literature selection, science mapping analysis, and a qualitative discussion based on 444 published
766 articles related to DT applications in project O&M. It was found that the number of published
767 articles in the studied domain significantly increased from 2020, indicating that the attention paid
768 to DT applications in project O&M from researcher and practitioners. Besides, there were 78
769 relevant peer-reviewed journals selected, finding that *Automation in Construction*, *Journal of*
770 *Manufacturing Systems*, and *International Journal of Computer Integrated Manufacturing*
771 contributed to the largest number (i.e., 11.71%) of published articles. Subsequently, the author's
772 keywords in DT applications in project O&M were identified by keyword co-occurrence analysis,
773 which includes BIM, human-robotic collaboration, and model-based system engineering. Some
774 keywords are related to basic technology utilization in new scenarios research like smart city, smart
775 manufacturing, and modular construction. Other emerging technology topics are still popular like
776 AI, blockchain technology, sensors, BIM, etc. These results suggest that experts concentrate more
777 on information integration issues because emerging technology can strongly change and facilitate
778 information management methods. In addition, it was found from the document analysis that the
779 most often referenced articles combining DT with other emerging technologies—such as
780 reinforcement learning, deep learning, transfer learning, and ML—remain a key area for enhancing

781 the time latency and data accuracy problems of DT, as well as provided as a simulation model
782 training platform.

783 For the qualitative discussion, mainstream research topics, research gaps, and future research
784 trends and challenges of DT application in project O&M were identified. The mainstream research
785 topics include (1) DT-based artificial intelligence (AI) technology for project O&M, (2) DT-
786 enabled smart city and sustainability, (3) DT applications for project asset management, (4)
787 Blockchain-integrated DT for project O&M, and (5) DT for advanced project management. Based
788 on the identified mainstream research topics, the research gaps and future research directions were
789 discussed. Lastly, a research framework was proposed based on linking the mainstream topics,
790 research gaps, and future research trends. Research into the application of DT to the construction
791 sector is of utmost importance because of the wide range of stakeholders, significant amount of
792 capital, lengthy operation and maintenance cycles, and the influence of construction hazards. This
793 study contributes to identifying significant journals, areas, and publications that have dealt with
794 connected subjects, adding to the body of knowledge and potential future research directions for
795 the application of DT in the construction sector.

796

797 **6. Limitations and future research works**

798 There are several limitations of this review study. First, the number of available articles pertinent
799 to the studied topic is not enough, and the topic is comparatively new, which causes the
800 scientometric analysis not to have an adequate sample size, and the total link strength of most
801 articles was too short which affected the influential research of authors, articles, journals,
802 countries/regions. Second, detailed technical analyses of specific emerging technologies in O&M
803 scenarios are lacking, providing only an overall theoretical framework and view of operations for
804 the phase. Third, the collection of articles ends in January 2024 from the Scopus database, and the
805 selection of the articles was limited to only English, thus, journal articles in other languages were
806 excluded. These exclusion criteria could cause a decline in annual publishing numbers and trend
807 graphs starting in 2024. Consequently, time restrictions are most likely to affect this phenomenon.
808 However, the detailed qualitative discussion combined with the science mapping analysis
809 supported by the existing insights and opinions by measurement analysis which increase the
810 science and reliability of a regular literature review.

811 To address the limitations, future studies should include more in-depth explanations of the
812 significance of new technologies related to DT systems for project O&M, along with illustrative
813 examples. Next, incorporating specific case studies would have improved the technical
814 explanations' realism and concreteness, making them simpler to comprehend and accept. In
815 addition, project managers must update their digital expertise and assist the DT technology to
816 realize staff career and job information management, workforce, and other resource deployments.
817 Moreover, standards for security assurance and standardized information communication practices
818 should also be provided by the integration and configuration of various technologies. Furthermore,
819 the advantages of DT for project O&M still require development with government policy support,
820 broad industry acceptance, and company adoption capacity. Lastly, to increase the sample size of
821 the included articles, further studies should include articles published in other languages and
822 databases.

823

824 **Data availability statement**

825 The datasets used in this study are available from the corresponding author upon request.

826 **Declaration of competing interest**

827 None

828

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