



SCIENTOMETRIC REVIEW OF CONSTRUCTION PROJECT SCHEDULE STUDIES: TRENDS, GAPS AND POTENTIAL RESEARCH AREAS

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Abstract. Scheduling plays a fundamental role in construction projects' success and thus has drawn attention from both academic researchers and industry practitioners. A large number of research articles tend to solve emerging challenges in construction project schedule (CPS). Therefore, there is a strong need of systematic review on existing studies. In this study, a total of 332 articles were retrieved from Scopus database using title, abstract and keywords with respect to CPS and filtered by document type, language type and abstract content. In particular, science mapping approach was adopted to analyse selected journal articles. These articles were examined using three sequential processes, including bibliometric search, scientometric analysis, and in-depth qualitative discussion. It could demonstrate the most influential journals, researchers, published articles, and active countries/regions in this area. In addition, major CPS knowledge areas were identified and summarized as CPS constructability, applications of variety of CPS methods, CPS optimization models and algorithms, identification and quantification of schedule risks and uncertainties, CPS performance management, and adopting new emerging CPS technologies and methods. Furthermore, knowledge gaps and future potential research directions were also discussed in detail. Finally, a comprehensive CPS framework was proposed as a sound reference in future research.

Keywords: construction project schedule (CPS), scheduling methods, schedule uncertainties, resource-constrained scheduling, schedule optimization, scientometric analysis, bibliometric.

Introduction

Developing a robust construction project schedule (CPS) is one of the major factors towards of construction projects' success. However, meeting project schedule is being challenged due to a number of factors such as project schedule reliability (Gannon et al., 2012), choice of CPS methods (Abou-Ibrahim et al., 2019; Al Nasser et al., 2016; Xu et al., 2018), complexity of projects (Abou-Ibrahim et al., 2019), probable existence of schedule risks and risk prediction capability (Choudhry et al., 2014; Liu et al., 2015; Luu et al., 2009; Shen et al., 2017; Soto Ramírez et al., 2018), project finance (Larsen et al., 2016), extent of project team collaboration (Sinesilassie et al., 2017) and other factors. Previous studies (Bakry et al., 2016; Ghodousi et al., 2017; Russell et al., 2013, 2014) indicated that

providing management reserve (such as providing an allowable buffer time) helps to manage prominent risks and uncertainties. As stated in Poshdar et al. (2016, 2018), reasonable amount of buffer can be estimated using early start and finish time of each activity. Moreover, increased emphasis on planning effort by practitioners contributes to develop reliable CPS (Lekshmi & Unnikrishnan, 2018; Lines et al., 2015). Recently, researchers have been focusing on emerging technologies such as four dimensional building information modeling (4D-BIM) which links 3D geometrical design with project schedule (Chan et al., 2015).

Scholars have published a large number of papers in worldwide recognized peer-reviewed journals in wide areas of CPS research including advancement of CPS meth-

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ods (e.g., BIM) (Mirzaei et al., 2018; Wang & Rezazadeh Azar, 2019), CPS optimization (e.g., resource constrained schedule optimization) (García-Nieves et al., 2018; Giran et al., 2017), time-cost-trade off analysis and CPS crashing (Gwak et al., 2016; Moselhi & Alshibani, 2013), CPS control (Tang et al., 2014) and schedule control. Some researchers have conducted reviews on specific CPS topics, which are summarized in Table 1. For instance, Ballesteros-Pérez et al. (2019) have examined non-linear time and cost trade-off analysis using previous literature and developed a genetic algorithm model with combined application of activity crashing and fast-tracking for CPS compression of construction projects. Yang (2017) has conducted a review that the total float was the most commonly used for schedule management which may lead to non-realistic management of floats and provide strategies for classifying and understanding of floats as an alternative float management. As stated in Al Nasser et al. (2016), using unstructured review, selection of CPS methods or combination of methods needs to consider multidimensional aspects such as usability, size and complexity of a project. Bragadin and Kähkönen (2016) have developed a schedule quality assessment model based on identified

schedule attributes such as construction process, schedule mechanics, and CPS control process. Furthermore, Vidhyasri and Sivagamasundari (2017) have examined 40 relevant articles to identify factors affecting CPS including planning and controlling system, climate conditions, and resource availability.

Using the guide developed by Dochy (2006), review papers can be classified as “*integrative*”, “*critique*” and “*state-of-the-art*”:

1. *Integrative* reviews provide comprehensive summaries on outcome of related previous papers.
2. *Critique* provides critical judgments based on subject matter of papers.
3. *State-of-the-art* considers the most current research papers.

As shown in Table 1, lack of systematically peer-reviewed publications has been noticed. Despite previous peer-reviewed papers’ contribution on specific CPS topics, the following limitations are noticed: (1) they addressed specific topics that may had biased tendency due to manual review; (2) all of them were integrative and critique reviews that have a tendency of subjectivity and thus may be lack of new perspectives; (3) no peer-reviewed papers

Table 1. Summary of previous peer-reviewed review articles

Researcher (1)	Review type (2)	Review method and tools (3)	Study period (4)	Database sources (5)	Focus of study (6)	Study finding (7)
Ballesteros-Pérez et al. (2019)	Integrative	Manual	1961–2019	–	Fast tracking based schedule compression for projects having non-linear time and cost relationship	Fast tracking model for both renewable and non-renewable resources
Zareei (2018)	Integrative	Manual	–	–	CPS methods for large scale biogas projects and having interdependent activities with varying duration	Adopted typical network diagram based CPM method
Yao et al. (2018)	Integrative	Manual	–	–	Optimization of multiple objective scheduling for construction projects	Heuristic, metaheuristic and mathematical models
Yang (2017)	Critique	Manual	–	–	Strategies for classifying and understanding of float management	Five managerial essentials and three proactive strategies
Vidhyasri and Sivagamasundari (2017)	Critique	Manual	1992–2017	40 papers with un specified database	Factors affecting performance of CPS	Identifies top factors such as resource availability, weather, and regulations.
Al Nasser et al. (2016)	Integrative	Manual	1956–2016	–	Choice of CPS methods to enhance success of construction projects	Provide taxonomy of CPS methods
Bragadin and Kähkönen (2016)	integrative	Manual	–	–	CPS health assessment using schedule requirements	CPS assessment model
Faghihi et al. (2015)	Integrative	Manual	1985–2014	–	Artificial intelligence for CPS automation	Model-based, expert systems, case-based reasoning, genetic algorithm
Zhou et al. (2013)	Integrative	Manual	–	–	Optimization of CPS	Heuristic methods CPS optimization
Yahya and Mohamad (2011)	Critique	Manual	–	–	Enabling rapid construction via lean principle	Eight enablers to integrate with lean

were systematically and logically structured. Thus, it is imperative to systematically review the state of art of CPS researches in order to understand the status of research trends, identify research gaps, provide future directions and identify accepted practices as well.

Construction project schedule is one of the requirements of successful construction project management. It plays a significant role in determining the flexibility with the available resources and complex precedence relationships, and thus becomes a substantial challenge worldwide. Though a lot of academic researchers and industry practitioners paid attentions to it in the last few decades, there still exist a large number of project failure associated with project scheduling. Researchers have investigated problems concerning with CPS's reliability, methods, causing factors and risks, resource assignment and CPS emerging technologies. However, there is a strong need of systematic review on existing studies of last decade (2009–2019). Thus, a comprehensive review state-of-the-art of CPS would be relevant to provide holistic picture for practitioners and researchers via systematic and science mapping approach.

This research paper aims to review CPS papers to serve as a basis for practitioners and researchers for further discussions on both undergoing and new research topics. In order to achieve the objectives, this paper is structured as follows. Section 1 presents methodology about science mapping approach and further qualitative

discussion. Section 2 states initial results through science mapping approach. Section 3 shows the extended science mapping approach to further discuss the research topics, research gaps, and trends in CPS. Finally, conclusions are drawn.

1. Research methodology

In order to conduct a comprehensive review, it is crucial to follow a systematic methodology. This study adopted science mapping approach to develop a bibliometric map based on the Scopus database sources between 2009 and 2019 regarding construction project scheduling. Figure 1 shows the steps of the review framework and processes, which include a bibliometric search, a scientometric data analysis and a detailed discussion of analysis results.

1.1. Bibliometric search data collection

A bibliometric search provides sufficient information of necessary records (Bankar & Lihitkar, 2019). Google Scholar (GS), Scopus and Web of Science (WoS) have been widespread database sources (Bakkalbasi et al., 2006). GS found the largest percentage of citations, and indexed any document that has an academic structure (Martín-Martín et al., 2018). This usually leads to inconsistent result (Falagas et al., 2008). GS might also lead to double citation counts, reporting higher citations than the reality when

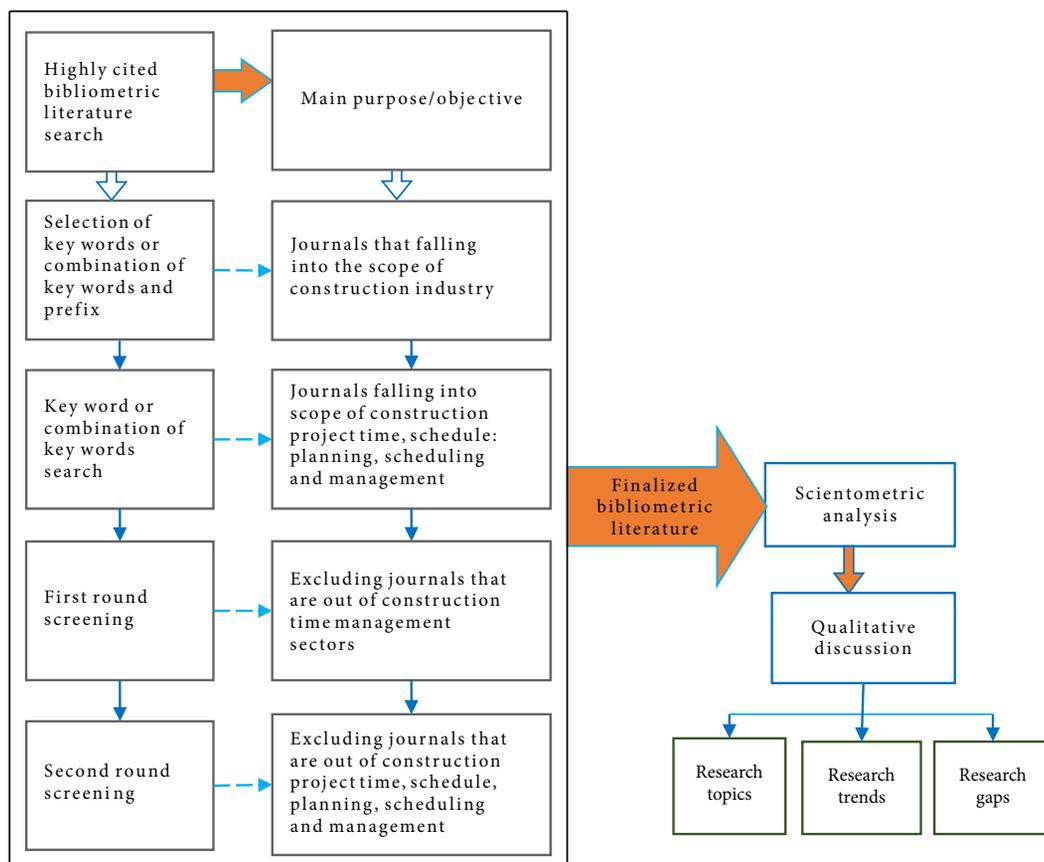


Figure 1. The methodological process

two or more versions of same journal articles are found online (Jacsó, 2005). Accordingly, citations found in GS need some substantial process to identify and clarify between double citation counts and homonyms (documents that look alike). Thus, incidental mistakes on citations in GS are subjected to errors. These might have major impact on citation metrics which lead to inconsistent result. However, Scopus indexed journal articles have no such type of problems, and articles are not subjected to double citation counting. Thus, citations of documents from Scopus are more reliable than GS. Meho and Yang (2006) indicated that citations in GS usually helpful for seeking an academic position or research grants. Moreover, if there is no problem of double citation counts in GS, the higher number of citations in this database could not change the ranking of most cited documents compared to Scopus (Meho & Yang, 2006). For instance, in this study, research conducted by Luu et al. (2009) (“Quantification of schedule risks using belief network in construction projects”) has been ranked first in both Scopus and GS with 98 and 209 citations respectively. Similarly, an article conducted by Li and Zhang (2013) has been ranked last in both Scopus and GS with 35 and 17 citations. It is more reliable that citations in Scopus are consistent to rank documents in comparison to GS.

Moreover, Scopus and WoS have higher academic contribution by focusing on selective scientific journal articles (Aghaei Chadegani et al., 2013; Martín-Martín et al., 2018). In addition to this, GS have less coverage in engineering and natural sciences compared to Scopus and WoS; and there exist a higher percentage of overlap with GS in this subject category (Martín-Martín et al., 2018). In this study, 1224 and 1152 journal articles have been searched from GS and Scopus respectively. There are a total of 1092 articles are the same from the both, which accounts for 94.8% (1092/1152). After the preliminary screening, more papers fall in the CPS from Scopus rather than that from GS. In particular, a number of documents indexed by GS are not open accessed. Moreover, Scopus covers a wider range of more recent publications than web of science (Aghaei Chadegani et al., 2013). Thus, to balance the academic contribution and coverage, this study used bibliometric search of the Scopus database engine, which is one of the largest source of peer-reviewed literature comprising of the largest number of citations and abstracts (Bakkalbasi et al., 2006; Martín-Martín et al., 2018). Furthermore, open access articles that are found only in GS have been retrieved and used in this study when they are necessary. In order to do a comprehensive review, the following data collection procedures have been followed: (1) literature search for highly-cited research papers on the topic of construction planning and scheduling to determine the related searching key words. Many search items, in this stage, include publications of other sectors than construction due to the reason that the term “planning” and “scheduling” are commonly used in many domain areas. Also, most of the search items in the

construction industry cover large domain area instead of more specific to project schedule; (2) combining the most important pre-fix search terms “construction project” (since construction schedules are usually prepared in project basis) with the term “planning” or “scheduling” or both after reviewing preliminary research publications in the first procedure to search CPS related literature; and (3) bibliometric search was conducted using combination of title, abstract and keyword search code in the domain of CPS as follows: TITLE-ABS-KEY (“construction project schedule”) OR TITLE-ABS-KEY (“construction project planning”) OR TITLE-ABS-KEY (“construction project scheduling”) OR TITLE-ABS-KEY (“construction project “planning and scheduling”) OR TITLE-ABS KEY (“construction project time”) OR TITLE-ABS-KEY (“construction project schedule” management) OR TITLE-ABS KEY (“construction project time” management).

During the searching, English publication language was chosen and journal articles were selected as journal papers, which usually provide more theoretic contributions compared to others including conference papers. The publication articles chosen were from the years 2009 to 2019 since wide attention of CPS was given in 2009 as shown in the Figure 2. In order to exclude articles having no related abstract with CPS, manual review was conducted to identify domain areas of CPS papers before going on to analyze the retrieved bibliographic data. After careful and repetitive manual review of papers, 332 research articles were ultimately retrieved for further scientometric analysis.

1.2. Scientometric data analysis

Manual review of such a number of research publications is time consuming and less reliable due to its subjectivity. In this paper, scientometric data analysis was adopted for systematic review of bibliometric literatures (Cobo et al., 2011). *VOSViewer*, a scientometric analysis software tool, was chosen to construct and analyze bibliometric network of science due to its special mining and visualization feature for large amounts of textual data (Van Eck & Waltman, 2011). Furthermore, there are a number of extant studies that have used *VOSViewer* in the field of construction management, i.e., construction waste management (Jin et al., 2019a), construction safety (Jin et al., 2019b) and building information modeling (He et al., 2017). *VOSViewer* scientometric analysis is a distance-based bibliometric network visualization between nodes to identify the extent of the relation on the network (Van Eck & Waltman, 2014). For the analysis, retrieved bibliographic publications were exported with all essential information records including titles, abstracts, and keywords, number of citations, sources, authors, institutions and references which are relevant for creation of bibliometric network. In this study, *VOSViewer* was applied to: (1) load 332 journal articles (i.e., an input data) with CSC format file that have been downloaded from Scopus; (2) create, visualize and analyze bibliometric networks to identify influential

researchers, publication papers, and journals; and (3) map and analyze the trends of research topics using bibliometric network of words in the domain of CPS.

The analysis is done based on the output of the data mining software (i.e., *VOSViewer*), which have been used in many literature reviewing articles. Thus, inspired by the capability of *VOSViewer* to extract the textual information of bibliometric data and to construct bibliometric network (Van Eck & Waltman, 2011), 332 journal articles (inputs) are loaded into *VOSViewer*. The textual information of these articles is extracted and visualized through bibliometric networks such as “citation-documents”, “co-occurrence-author key words”, and “citation-authors” to identify the most cited articles, studied topics, and influential authors respectively in the domain of CPS during the last decade. Articles’ different information could be visualized in different bibliometric networks when the textual information of journal articles has a strong relationship with the subject matter. For instance, journal articles with high citation number might be frequently studied topic or they might be from an influential country or region in the area of that topic. Thus, they could be cited in various sections of this paper and might be used redundantly if they are influential with regard to various aspects. In this regard, journal articles from 332 input data might not be used and cited in this paper if their textual information is extracted and visualized in the bibliometric network due to their limitation regarding the subject matter. Thus, among the 332 journal articles, only relevant articles related to the subject matter are used and cited to support the discussion of CPS topics and subtopics where they are necessary.

1.3. Qualitative discussion

Detailed qualitative discussion was also provided based on scientometric analysis to summarize the mainstream of research, identify existed research gaps and propose future directions in CPS research. Finally, comprehensive framework connecting mainstream research trends with future research areas is proposed to create clear understanding for researchers.

2. Research analysis result

2.1. An overview of a bibliometric literature data

The number of papers published from 2009 to 2019 in the domain of CPS is shown in Figure 2. The number of published papers was incomplete for 2019 as the data was collected up to April 29, 2019. However, the number of research publications shows a significant increase trend of 9 papers in 2010 to 55 papers in 2018. It indicates that the domain of CPS highly attracted the attention of many researchers in the last decade.

2.2. Main influential journals

The main influential journals in the area of CPS are shown in Figure 3 and Table 2. These influential journals were

visualized and analysed using ‘citation-documents’ bibliometric network in *VOSViewer*. For each journal, minimum of three research papers and 30 citations were set for further analysis. Accordingly, 11 out of 111 journals met the minimum thresholds. Figure 3 shows the citation bibliometric network, in which nodes are journals and lines from one to another shows their citation relationships. It is noted that the names of journals may not be shown fully in *VOSViewer* and the omitted information can be seen in Table 2.

As shown in Figure 3, the size and font of journal nodes are clustered differently based on their number of published papers and citations. Journals with high number of papers and citations are larger in node and font size and journals having a mutual citation are strongly related each other (Van Eck & Waltman, 2014). For instance, *Journal of Construction Engineering and Management*, *Automation in Construction*, and *Journal of Civil Engineering and Management* were the major influential journals. Other quantitative measurements of journals are summarized in Table 2.

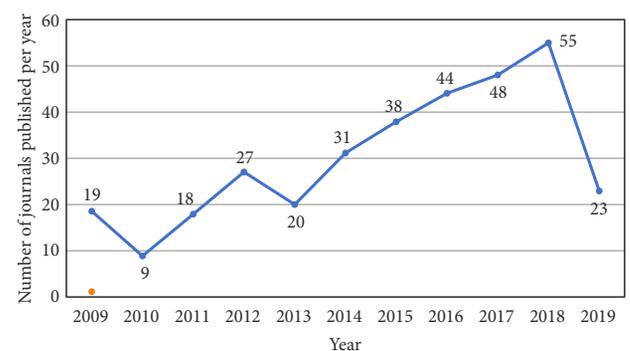


Figure 2. The trend of number of published papers per year from year 2009 to 2019

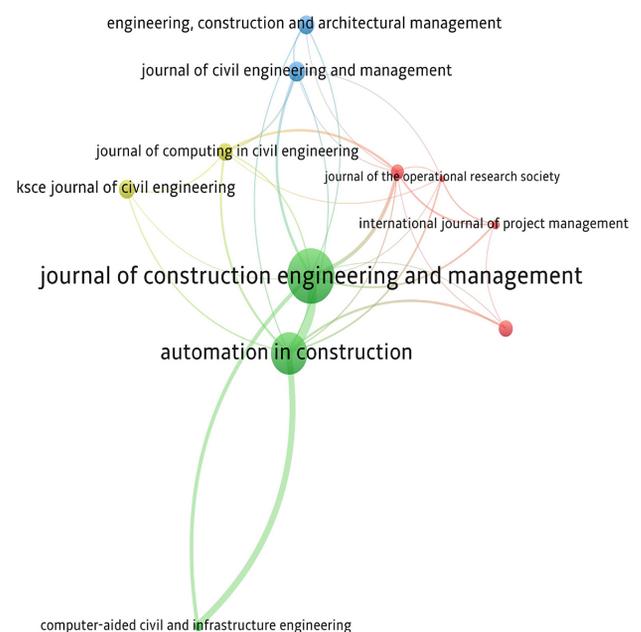


Figure 3. Science mapping of main influential journals in the domain of CPS

Table 2. Summary of measurements of journals publishing CPS research

Journal name	Number of publications	Citation	Average citation	Average normal citation
Journal of Construction Engineering and Management	54	668	13	1.21
Automation in Construction	37	598	16	2.13
Journal of Civil Engineering and Management	13	73	6	0.64
Engineering, Construction and Architectural Management	12	72	6	0.72
KSCE Journal of Civil Engineering	12	31	3	0.46
Journal of Computing in Civil Engineering	11	214	20	2.83
Journal of Management in Engineering	10	101	10	1.56
Construction Management and Economics	9	75	9	0.75
Computer-Aided Civil and Infrastructure Engineering	5	35	7	1.88
International Journal of Project Management	5	224	45	2.02
Journal of the Operational Research Society	3	59	20	1.45

Total citations, average citation, and average number of normal citation are used to measure and quantify the influence of journals in CPS researches. In particular, average number of normal citation is equal to dividing the total citation number to average citation number per year. The normalization adjusts the misconception that older articles gain ample time to have citations than more current publications (Van Eck & Waltman, 2014). It is also used in subsequent tables to measure the influence of researchers, papers, author keywords, or countries in CPS research. Journals having small number of research papers publications may have highest average citation and average normal citations. For instance, *Journal of Computing in Civil Engineering* is less productive in number of publications (i.e., 11 papers) compared to *Journal of Construction Engineering and Management* (i.e., 54 papers), but received more average citations and normal citations. *Journal of Construction Engineering and Management*, and *Automation in Construction* have highest citations showing that they are the most influential journals in terms of productivity and contribution with an intermediate yearly influence of average citation and normal citation. Moreover, *Journal of Civil Engineering and Management*, and *Engineering, Construction and Architectural Management* are also the best journals contributing for publishing research papers but least yearly influence in terms of average citation and average normal citation in the domain of CPS. There are also other journals having highest yearly average citation but a less influence in production of research papers and contribution for research community such as *International Journal of Project Management* and *Journal of the Operational Research Society*.

2.3. Science mapping of co-occurrence of key words

The key words in the bibliometric network indicate major outlines of previous researches and depict main topics of intellectual teams and organizations (Van Eck & Waltman, 2014). This study adopted science mapping of co-occurrence of words using “fractional counting” and “author

key words” in *VOSViewer* to create bibliometric network (Hosseini et al., 2018). The minimum occurrence of major key words was set 3 times. During the initial result, 105 key words of out of 891 met the threshold, from these many general words were ignored, for example, “construction”, “construction industry”, “construction projects”, “construction planning”, “project scheduling”, “scheduling”, “construction management”. Moreover, some other key words with similar semantic connotations (i.e., “critical path method” versus “CPM”, “building information modelling” versus “BIM”) and words having exact meaning, but plurality difference (i.e., “buffer” versus “buffers”, “risk” versus “risks” and “genetic algorithm” versus “genetic algorithms”) are selected with higher occurrence value. Lastly, a total of 46 keywords were chosen as shown in Figure 4 and Table 3. The detailed quantitative measurement of most 37 keywords is summarized in the Table 3.

Figure 4 shows the most frequently occurred keywords represented by different nodes. These include but not limited to optimization, critical path method, resource levelling, resource allocation, uncertainty, schedule risks, and genetic algorithm. These frequently keywords are clustered differently in node size, colour and font size and interconnected by lines based on their relatedness. Figure 4 presents that lean construction and last planner, optimization and resource levelling, building information modelling and labour productivity are strongly related to each other from similar cluster. Keywords can be also be strongly related from different clusters such as resource allocation and genetic algorithm. In general, the most major keywords that have been studied in the domain of CPS are categorized as follows:

- *Scheduling methods and tools*: CPS methods such as Critical Path Method (CPM), Line of Balance (LOB), linear scheduling, analytical methods and network analysis have been studied in previous studies (Abbondati et al., 2016; Ali & Elazouni, 2009; Alias & Ismail, 2012);
- *Schedule risks and uncertainties*: CPS risks and uncertainties have been studied by many researchers re-

Table 3. Summary of the most frequently occurred keywords in the domain of CPS research

Key words	Occurrence	Avg. year published
Optimization	30	2014
Critical path method	15	2015
Risk management	14	2015
Project planning	12	2014
Linear scheduling	9	2016
Constraint programming	9	2015
Genetic algorithms	9	2013
Resource allocation	9	2014
Uncertainty	9	2015
Monte Carlo simulation	8	2017
Resource levelling	8	2017
Risk analysis	7	2016
Lean construction	7	2015
Network analysis	7	2015
Last planner system	6	2016
Fuzzy sets	6	2013
Building information modelling	5	2016
Time-cost-trade off	5	2015
Schedule risks	5	2016
Multi-objective optimization	5	2015
Analytical techniques	4	2015
Repetitive construction projects	4	2015
Project planning and design	4	2015
Primavera	4	2017
Simulation models	3	2012
Crashing	3	2018
Fast tracking	3	2016
Forecasting	3	2013
Work flow	3	2017
Buffer	3	2014
Resource constrained project schedule	3	2013
Resource utilization	3	2014
Labor and personal issues	3	2014
Production control	3	2013
Project control	3	2017
Line of balance	3	2016
Linear programming	3	2017

Note: Keywords are listed based on their occurrence value from highest to the lowest.

lated to identification, cause-effect and their analysis (Choudhry et al., 2014; Shen et al., 2017; Xu et al., 2018);

- *Schedule optimization algorithms and models*: A number of CPS optimization algorithms have been studied to solve multiple CPS problems, either single objective or multiple objectives, with single or multiple

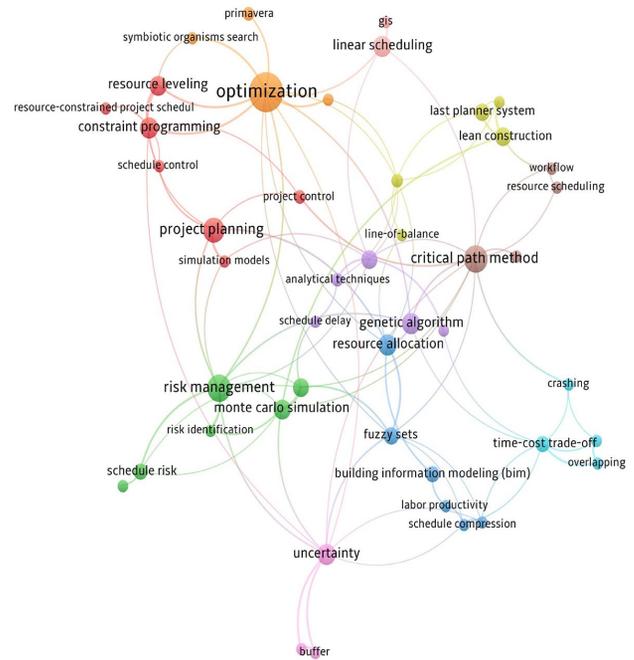


Figure 4. Bibliometric network of keywords

constraints such as genetic algorithm (Long & Ohsato, 2009; Zou et al., 2016), fuzzy stochastic network (Radziszewska-Zielina et al., 2017), Monte Carlo simulation (Indhu & Farhan, 2015), fuzzy mathematical models (Castro-Lacouture et al., 2009), constraint programming, linear programming, and symbiotic organisms search (Cheng et al., 2016);

- *CPS emerging technologies*: CPS emerging technologies have been examined in previous studies such as 4D-BIM (Mirzaei et al., 2018) to detect time-space problems, geographical information system (GIS) (Bansal & Pal, 2009) to identify logical error and missed activities, integration of BIM with primavera (Subramani & Ammai, 2018) to manage risks and BIM schedule (Chen et al., 2011; Ding et al., 2015) to assign crew and other resources;
- *Resource scheduling*: One of the most research areas in CPS is to solve resource constrained scheduling problems (Christodoulou et al., 2012; Giran et al., 2017). It includes but not limited to, resource allocation (Khanzadi et al., 2016), resource levelling (El-Rayes & Jun, 2009), resource utilization (Siu et al., 2016);
- *Lean project scheduling*: The application of lean principle (e.g. Last Planner System (LPS)) has been widely studied to improve performance of project scheduling. For example, (Yahya & Mohamad, 2011) studied how to apply lean for rapid work flow. The technique is applied to solve CPS constraints via team collaboration (Alsehaimi et al., 2014; Daniel et al., 2017; Kovvuri et al., 2016);
- *CPS planning and management*: CPS planning and controlling of CPS has been examined. For instance, early project planning is the basis of CPS (Al Nas-

- seri & Aulin, 2016; Larsen et al., 2018) and schedule monitoring using control charts (Băncescu, 2016);
- *Factors affecting schedule*: CPS Labour productivity (Huang et al., 2014), resources (Larsen et al., 2016), labour and personal issues, risks and uncertainties (Kavuma et al., 2019) are examined by previous studies as they are problems affecting project schedule; and
- *Time-cost-trade-off*: Time-cost trade-off has been one of the most CPS research areas focusing on reduction of project duration with an optimal cost. For example, schedule compression using fast-tracking (i.e., overlapping of activities) for both linear and non-linear projects (Ballesteros-Pérez et al., 2019) and crashing (Sonmez et al., 2016).

Table 3 shows the summary of quantitative measurements of keywords. Optimization, schedule delay, symbiotic genetic algorithms, and resource allocation have been attracting the attention of scholars. The average publication year of keywords shows their recentness in the domain of CPS. For instance, studies focusing on work flow (i.e., one of the lean principles) and building information models were mostly published recently in 2017 and 2016 respectively. The measurement result is consistence in terms of average normal citation received for production control (lean principle such as last planner), which is one of the most research theme in the areas of CPS.

2.4. Co-authorship analysis

Academic researchers usually collaborate with others to stimulate new ideas and to enhance their productivity (Hosseini et al., 2018). This research used “citation” and “author” to create and visualize co-authorship bibliometric network. Accordingly, the minimum number of research papers and the citations received for an author were fixed at three and 45 respectively in *VOSViewer*. Finally, 19 authors from 711 met the minimum thresholds. Figure 5 and Table 4 show the most influential authors who have been contributing in the areas of CPS. Most of the authors are clustered in different groups including Mohamed and EI-Rayes group. For example, Lucko and Cho as shown in the Figure 5 may be group members excluded due to the minimum threshold.

As shown in Table 4, authors are measured using the following five measurements: number of articles published, total citations received in Scopus, average publications per year, average citations and average normal citations. Dr. EI-Rayes is the most productive author, with having significant total citations compared to other researchers. It indicates that he is the most influential author in the last decade in CPS research. Other scholars such as Dr. Liu and Dr. Mohamed are collaborators and contributors as shown in Figure 5. Dr. Tran and Dr. Cheng are the new emerging scholars whose research publications were published in 2017 whereas Dr. EI-Rayes contributed to the research community starting in 2010 as shown in the Table 4. The average normalized citation measurements of

scholars shows the extent of authors’ inspiration per year in CPS research, for example, Dr. Love, Dr. Hegazy and Dr. Wang have been attracting scholars’ attention even though they were not productive in terms of the number of articles and total citations compared to other scholars.

2.5. Citation of articles

The most influential articles that have been published in Scopus journals were analysed using *VOSViewer* during the last decade. This study fixed the minimum number of citations at 30. A total of 23 research articles from 332 articles met the minimum threshold. Figure 6 shows these most influential articles that have been published in the areas of CPS.

The detail measurement of top 15 influential research articles including scholars, research titles, received citations and average normal citations are summarized in Table 5.

It shows the most influential research articles that received highest number of citation such as Luu et al. (2009), which has focused on quantification of schedule risks.

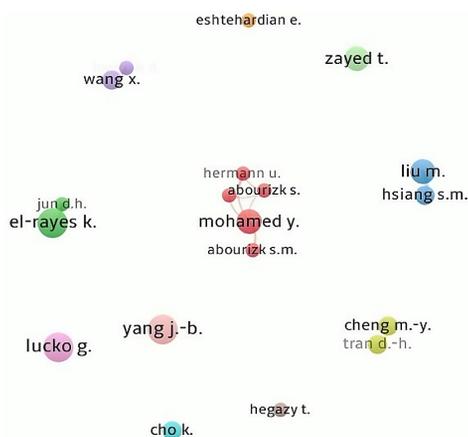


Figure 5. Co-authorship bibliometric representation
 Note: The full information of authors is summarized in Table 4.

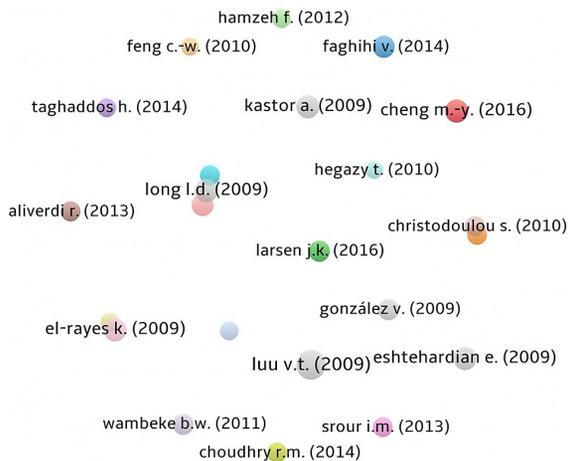


Figure 6. Mapping of most influential papers
 Note: the full information of most influential research article is summarized in the Table 5.

Table 4. Summary of measurements for co-authorship in CPS research

Scholar	Affiliation	Number of articles	Total citations	Average publication year	Average citations	Average normal citations
Khaled El-Rayes	University of Illinois at Urbana–Champaign	6	124	2013	21	0.9
Dho Heon Jun	University of Illinois at Urbana–Champaign	3	103	2010	34	1.6
X Wang	Curtin University	4	83	2015	21	3.6
Yasser Mohamed	University of Alberta	5	80	2014	16	1.4
Tarek Hegazy	University of Waterloo	3	71	2011	24	1.6
Duc Hoc Tran	Ho Chi Minh City University of Technology	4	68	2017	17	5.1
Ped Love	Curtin University	3	64	2014	21	1.9
Min Liu	North Carolina State University	5	62	2014	13	0.9
Min-Yuan Cheng	National Taiwan Univ. of Science and Technology	4	62	2017	16	3.9
Simon Hsiang	Texas Tech University	4	62	2013	16	1.1
Ehsan Eshtehardian	Iran University of Science and Technology	3	56	2015	19	0.8
Chung-Suk Cho	Chosun University Gwangju	4	54	2012	14	0.9
Hosein Taghaddos	University of Tehran	3	53	2015	18	1.8
Ulrich Hermann	PCL Industrial Management Inc	3	53	2015	18	1.8
Simaan AbouRizk	University of Alberta	3	53	2014	18	1.6

Note: Researchers in Table 4 are listed based on the number of total citations.

Table 5. List of publications with the highest impact in CPS

Scholar	Title	Total citation	Average normal citations
Luu et al. (2009)	Quantifying schedule risk in construction projects using Bayesian belief networks	98	3.24
El-Rayes and Jun (2009)	Optimizing resource levelling in construction projects	55	1.82
Eshtehardian et al. (2009)	Fuzzy-based MOGA approach to stochastic time-cost trade-off problem	53	1.8
Kastor and Sirakoulis (2009)	The effectiveness of resource levelling tools for Resource Constraint Project Scheduling Problem	50	1.65
Long and Ohsato (2009)	A genetic algorithm-based method for scheduling repetitive construction projects	50	1.65
Faghihi et al. (2014)	Construction scheduling using genetic algorithm based on building information model	45	3.92
Wambeke et al. (2011)	Causes of variation in construction project task starting times and duration	44	2.70
González et al. (2009)	Multi-objective design of Work-In-Process buffer for scheduling repetitive building projects	43	1.4
Larsen et al. (2016)	Factors Affecting Schedule Delay, Cost Overrun, and Quality Level in Public Construction Projects	41	7.61
Christodoulou (2010)	Scheduling resource-constrained projects with ant colony optimization artificial agents	38	3.2
Zhou et al. (2013)	A review of methods and algorithms for optimizing construction scheduling	38	2.8
Choudhry et al. (2014)	Cost and schedule risk analysis of bridge construction in Pakistan: Establishing risk guidelines	37	3.2
Aliverdi et al. (2013)	Monitoring project duration and cost in a construction project by applying statistical quality control charts	36	2.65
Li and Zhang (2013)	Ant colony optimization-based multi-mode scheduling under renewable and non-renewable resource constraints	35	2.57

Note: Research articles in Table 5 are listed based on the number of publications.

Luu et al. (2009) established cause-effect relationship between schedule risks to build Bayesian Belief Networks, an artificial intelligence tool, to represent probabilistic occurrence of variables using node and arrows to show the relationship between variables (McCabe et al., 1998). The result shows that material shortages, financial difficulties and lack of experience were the top ranked factors (Luu et al., 2009). These schedule factors are consistent with the study conducted by Choudhry et al. (2014) in Pakistan, which is ranked third article. El-Rayes and Jun (2009) and Eshtehardian et al. (2009) are the second and third ranked articles having highest total citations in the last decade respectively. The former article focused on resource levelling optimization using genetic algorithms. Eshtehardian et al. (2009) used fuzzy numbers to develop an optimized time-cost-trade off via multi-objective genetic algorithms to determine the level of risk acceptance. Kastor and Sirakoulis (2009) is ranked fourth and investigated an effectiveness of resource scheduling software packages (i.e., Primavera, Microsoft Project) for construction projects having resource constraints and activity priority constraints. In general, the following research themes were the focus areas of most influential research articles but not limited to:

- Identifying causes of variation affecting activities' planned starting time and duration, including response delay regarding clarification, quality of documents, weather condition, experience (Wambeke et al., 2011) and factors affecting schedule delay such as lack of finance (Larsen et al., 2016);

- Application of buffer time to manage probable occurrence of variations (González et al., 2009);
- Using ant colony artificial neural networks for both resource constraint and non-constraint projects (Chan et al., 2015; Christodoulou, 2010);
- Resource constrained repetitive project scheduling using genetic algorithms (Long & Ohsato, 2009) and using variety of skill workers to optimize schedule of linear projects; and
- LPS and employing new scheduling methods including BIM with genetic algorithm (Faghihi et al., 2014) and quantifying schedule risks (Luu et al., 2009).
- Adopting new scheduling methods such as BIM with genetic algorithm (Faghihi et al., 2014) and quantifying schedule risks (Luu et al., 2009) have a highest normal citation indicating that they attract the attention of research community. This indicates that one further research direction may investigate in schedule risk identification, quantification and prediction in relation to new schedule technologies.

2.6. Active countries/regions in CPS research

Table 6 and Figure 7 indicate the most influential countries/regions that have been active in the domain of CPS research. In *VOSViewer* scientometric analysis, the countries/regions with the minimum number of 4 articles and 45 citations were included. A total of 18 countries out of 58 met the threshold. In the bibliometric network, both

Table 6. Summary of most influential countries/regions in CPS research

Country/Region	Number of documents	Number of Citation	Average publication year	Average citations	Average normal citations
United States	56	692	2014	13	1.21
India	33	107	2016	4	0.5
Canada	31	309	2015	10	1.2
China	30	216	2016	8	1.2
South Korea	25	265	2015	11	0.9
Iran	25	157	2016	7	1.0
Taiwan	23	313	2014	14	1.6
Poland	18	102	2016	6	0.8
Australia	14	126	2016	9	1.5
United Kingdom	13	212	2014	16	2.9
Turkey	12	53	2015	5	0.6
Saudi Arabia	11	103	2016	10	1.0
Hong Kong	9	115	2015	13	1.7
Vietnam	7	173	2015	25	2.5
Egypt	7	58	2014	9	0.8
Greece	5	73	2013	15	0.8
Cyprus	4	47	2012	12	0.8
Lebanon	4	67	2016	17	1.3

Note: Countries/regions in Table 6 are listed based on the number of documents.

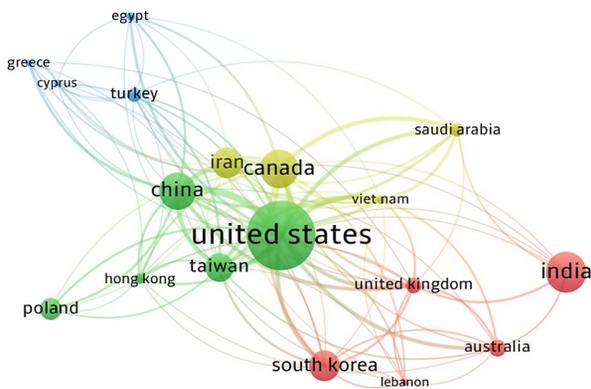


Figure 7. Science mapping of most influential countries/regions

developed and developing countries/regions contributed to the research community in CPS research including United States, Canada, China, Taiwan, Poland and India. Figure 7 and Table 6 shows the summary of quantitative measurement of major countries/regions. In Figure 7, countries/regions are represented by nodes and connected by lines, which indicates mutual citations of research papers within different countries. It is shown that United States, Canada, India, China and South Korea have larger node size and font size compared to other countries/regions. For example, many research articles regarding optimization of resource constrained schedules have been published in United States (El-Rayes & Jun, 2009), Taiwan (Hsie et al., 2009), Poland (Jaskowski & Biruk, 2018; Roslon & Kulejewski, 2019), China (Ma & Xu, 2014), India (Tiwari & Johari, 2015) and Iran (Kasravi et al., 2019). A summary of quantitative measurements of influential countries/regions in terms of number of articles published, citations received, average citation and normal citations are presented in the Table 6.

Researchers from United States are ranked first in terms of number of published articles and total citations followed by India, Canada and China. Much research has been done in United States related to resource scheduling such as optimization of time-cost-trade off (Ashuri & Tavakolan, 2012) and resource requirement fluctuation model (El-Rayes & Jun, 2009). Developing countries including China and India also conducted research related to time-cost-trade off problems (Zou et al., 2015), ant colony schedule-cost optimization (Li et al., 2013) and resource allocation improvement strategy (Zou et al., 2014). However, countries from Africa have less contribution in CPS research except Egypt which published articles on schedule optimization using swarm particle (Elbeltagi et al., 2016). This result is consistent as most of the construction projects are delayed in African countries. For example, projects in Ethiopia experienced 61–80% schedule slippage due to schedule management problems (Gebrehwet & Luo, 2019). Moreover, developed countries/regions such as United Kingdom, Austria, Taiwan, and United States, and Hong Kong have received a highest average normalized citation indicating that they have noticeable contribution for growth of CPS research.

3. Qualitative discussion

Based on scientometric analysis, the authors conducted detailed discussion to provide summary of major research topics, identify research gaps, and propose research framework for future research directions. The detailed discussions are illustrated in the following subsections.

3.1. Main research topics in CPS

CPS research covered a wide range of construction and civil engineering project management including project time, cost, quality, risks and human resource management. Based on the type and objectives of study, a variety of research methods and combination of them could be employed such as questionnaire, interview or questionnaire and interview together (Luu et al., 2009), review of existing practices (Al Nasser et al., 2016), simulation models such as Monte Carlo simulation (Choudhry et al., 2014), mathematical models such as fuzzy sets (Castro-Lacouture et al., 2009) or genetic algorithms for resource scheduling e.g. (Lin & Hsiau, 2010; Suresh et al., 2011). In this study, the main research topics of CPS during the last decade are categorized as follows.

3.1.1. Application of scheduling methods

Analysing the suitability, usability, benefit of CPS methods, and users' experience provides basis for implementation (Al Nasser et al., 2016). Projects that requires only milestone information may be scheduled using bar chart (Kim, 2012), projects having uncertainty can be scheduled using Program Evaluation Review Technique (PERT) (Hsiau & Lin, 2009) or combination of CPM with fuzzy set activity representation (Ökmen & Öztaş, 2014). Other most influential input for the selection of scheduling methods is nature of projects: linear, repetitive or conventional. For instance, linear activities (i.e., activities that are completed as they progress along their path) could be scheduled using Linear Scheduling Method (LSM) (Bonnal et al., 2013) whereas LOB could be used for small number of activities which have repetitive nature (Ali & Elazouni, 2009; Tomar & Bansal, 2019). Bonnal et al. (2013) also proposed unified scheduling system by integrating LSM, LOB and precedence diagram method (PDM) for projects having mixture of different kinds of activities. Furthermore, Tomar and Bansal (2019) also developed combination of CPM with LOB for projects having multi-characteristic nature. However, current available scheduling methods could not achieve collaboration of stakeholders due to the nature of multiparty involvement agreement. Yahya and Mohamad (2011) identified lean principles to reduce non-value adding activities through team collaboration's planning effort. Similarly, Kovvuri et al. (2016) also found that last planner benefits construction time reduction, quality, cost improvement, build trust, and increase coordination. Moreover, scholars have been continuously developing emerging scheduling methods such as Mahalingam et al. (2015) adopted lean principles with BIM to enhance CPS

performance via team collaboration. At the same time, CPS risks could be detected and identified to prevent the occurrences of time interruption during execution (Mirzaei et al., 2018).

3.1.2. Resource constraint scheduling

Renewable resources could be deployed based on activities' resource consumption so that resources are continuously supplied without constraints. In the real world, construction resources are limited so that they need to be utilized properly. Researchers developed resource optimization systems such as allocation and levelling using mathematical algorithms. For instance, Cheng et al. (2016) developed an application of discrete symbiotic organisms search to manage the fluctuation of temporally hiring and firing of staff and enables cut-off the peak resources requirement. Christodoulou (2010) applied ant colony artificial agent to determine effects of resource constrained using CPM whereas El-Rayes and Jun (2009) focused on resource levelling optimization via genetic algorithm by combining two assumptions: 1) quantifying the required resources during the low resource requirement period and redeploy additional resource for peak period; 2) estimate the wasted days due to fluctuation of resources requirement. Scholars, for example, Zhou et al. (2013) analysed and classified resource constrained CPS into the following categories: (1) mathematical which considers an objective function in the existence of constraints; (2) heuristic to optimize project specific schedule instead of its universality; and (3) metaheuristic for scheduling problems having natural processes such as Pareto optimality genetic algorithm. Genetic algorithm has been widely applied in time-cost trade-off analysis to shorten project duration (Amiri et al., 2017; Tran & Long, 2018) and time-cost-quality trade-off to enhance quality (Abd El Razeq et al., 2010). Al Haj and El-Sayegh (2015) used nonlinear-integer programming to shorten project duration and minimize project cost taking into total float time. Other studies, including but not limited to, Ezeldin and Soliman (2009) developed hybrid method that combined dynamic programming and genetic algorithm to maximize the efficiency of project time and cost under environmental uncertainty.

3.1.3. Identifying and quantifying schedule risks and uncertainties

As stated in Luu et al. (2009), material shortages, financial difficulties and lack of experience were identified as the top CPS risks. Scholars identified different CPS risks from different perspectives. For example, financial risk has been identified as the most serious CPS risk both in terms of effect and occurrence (Choudhry et al., 2014; Larsen et al., 2016). A Monte Carlo simulation technique conducted by Liu et al. (2015) indicated that geological risks have high level of risks on tunnel boring machine during design phase. Poor communication between resource planning units, inconsistent information flow in logistics planning system due to human errors probability, precast element delivery delay, and poor information between designer

and producer have been identified as CPS risks via Monte Carlo simulation in the production of prefabrication houses in Hong Kong (Shen et al., 2017). Furthermore, Xu et al. (2018) indicated that performing various sensitivity and uncertainty analysis helps for the precise prediction of unknown uncertainties and risks which may cause project delay during an execution of the projects.

3.1.4. Planning and controlling schedule

The alignment of planning and scheduling is an essential element of CPS in order to establish basis for controlling and evaluating of project performance. However, factors such as human, technological, organizational and management affect quality of project planning (Tesfaye et al., 2017). Organization's management practices determine how the projects are properly planned within an acceptable schedule practice. Lines et al. (2015) developed pre-contract model which was implemented between contractor and client prior to the contract award. The model showed that pre-contract planning reduced project cost by 54% and improved CPS by 70% with increased client satisfaction. Other studies such as Ryall et al. (2012) identified that LPS brings all stakeholders' effort to develop reliable CPS so that project's track can be effectively controlled. Hamzeh et al. (2012) identified factors affecting last planner system such as existence of non-compliance with LPS rules, and inadequate standardized practices. It has been noticed that CPS planning plays a significant role to develop robust schedule and control compared to planned CPS baseline. Moreover, selection of construction methods during the planning stage is also crucial for successful completion of projects. It does affect the activities, their durations and their work sequences. As a result, a key decision is required for the proper development of construction project in relation to construction methods (Thomas et al., 1990). In this process, scheduling is highly iterative process and demands the project team to examine various data sources and apply own experience to develop a set of efficient methods. Udaipurwala and Russell (2002) examined a case study using actual data via a simulation technique to choose better construction methods. Moreover, lean construction in which all stakeholders involve in the early stage of project with the support of industry practitioners to develop an integrated project schedule (Song et al., 2009).

3.2. Current research gaps within CPS

Identifying the forthcoming directions has become an extensive and accepted method to improve the current industry practice (Harty et al., 2007). In this paper, future research directions are identified by synthesizing the most cited previous studies together with the current project schedule requirements. Accordingly, two consecutive steps are used to do this. Firstly, it is necessary to conduct a comprehensive disclosure of the sets of information from previous studies including future directions stated by previous studies to get insights. During the synthesis of

most cited journal articles, the authors find a “space” to identify future directions. These are regarding gaps that need an improvement or new researches in relation to the current need of scheduling requirement, especially, artificial intelligence (AI) which can help to formulate model and automate tasks (Acemoglu & Restrepo, 2019). In this step, a number of issues have been listed and categorized including, but not limited to, project early planning, automated schedule performance evaluation, selection of CPS methods, dynamic multi-objective project schedule, project schedule data analytics, and other schedule applications such as BIM and AI-based schedule. Secondly, the authors conducted systematic searching of publications in other databases such as Google scholar and Web of Science using the identified gaps in the first step. Moreover, the most cited journal articles are synthesized using available evidences in conjunction with current scheduling requirements to improve the current industry practice using artificial intelligence techniques. Accordingly, previous studies lack the application of emerging digital technologies in order to improve project schedule performance. May et al. (2018) indicated that construction projects have been significantly affected by loss of productivity, which resulted an average of 20% schedule slippage due to lack of digitalisation. The application of full digitalization in construction projects can be effectively reduce schedule overruns by 10–15% (May et al., 2018). For this, the shift towards digital technologies in conjunction with schedule performance management will pave the way for successful implementation of projects. With that in mind, adopting digital technologies to develop construction project schedule is vital to enhance their schedule performance. For instance, BIM has been extensively used in the field of engineering, architecture and construction sectors to build the coordinated and systematic information platform to be accessed by project stakeholders (Succar, 2009). BIM has been still used to reduce project duration through collaboration and communication among stakeholders; and it remains to be favored within the research community in the future (Abdirad & Pishdad-Bozorgi, 2014; Eleftheriadis et al., 2017). Moreover, Müller and Bostrom (2016) indicated that the development of high-level machine intelligence will be raised by 50% around 2040–2050 while rising to 90% by 2075 in order to increase the consistency and precision of the results. With that in mind, adopting digital technologies and other artificial intelligence in construction projects is vital to enhance schedule performance. Finally, the following current research gaps are categorized for future studies even though wide areas of CPS were addressed in previous studies.

3.2.1. Front-end schedule reliability assessment

Front-end planning defined by Faniran et al. (2000) at an early stages of project when crucial and binding elements of project scopes are analysed and incorporated including project's feasibility and execution strategies. This requires an understanding of methods and tools, skills and knowledge, and experience about underlying concepts of

planning and scheduling prior to implementation of CPS. AlNasseri and Aulin (2015) indicated that an incorporation of detailed project scope during front-end planning significantly contributes for project success. For instance, Lines et al. (2015) has examined that project duration could be shorten by 54% and minimized cost by 70% if pre-contract planning is incorporated. However, front end planning effort is given less attention during the initial project schedule (AlNasseri & Aulin, 2015; Zwikael, 2009). The study carried by AlNasseri and Aulin (2015) indicated that recognition and adoption of front end planning in construction projects is limited and not sufficiently implemented. The verification of scope of works and identification of project schedule risks were poorly performed in early stage of project (Gannon et al., 2012; Petrochenko et al., 2018). Furthermore, study done by Bragadin and Kähkönen (2016) demonstrated that schedule quality assessment metrics are relevant in order to verify CPS deficiencies in early stages of projects despite the study has limitation in incorporation of CPS risks and perception of stakeholders. For this, the application of artificial intelligence can increase the credibility and precision of the results by automating the tasks (Agarwal et al., 2016). Wang et al. (2012) proposed artificial neural networks to predict project schedule based on projects' early planning information. The study indicated that reliable early planning contributes for an accurate prediction of project success using an automated artificial intelligence models. Thus, identifying the current practice of CPS early planning shortcomings and adequately developing project early planning for complex projects using artificial intelligence techniques such as Support Vector Machines requires further research on front end CPS to develop an integrative schedule review model at the earliest stage of the project.

3.2.2. Optimizing choice of scheduling methods

Many researchers investigated applications of different CPS methods. Al Nasseri et al. (2016) indicated that project size and complexity, usability and suitability, underlying basic concepts and theories of these methods, and project manager's capability are some of CPS methods selection parameters. Similar to the identification of the characteristics of each CPS method, projects' parameters and requirements have to be adequately defined to formulate the associated CPS method selection models. For this, manual selection of respective CPS methods for complex projects having a number of project parameters is challenging and time taking, and probably leads to imprecise results. For instance, Rakhshani et al. (2019) adopted deep learning models to provide solutions for enormous nasty optimization problems with an excessive parameters. With that in mind, adopting such kind of artificial intelligence models for selection of CPS methods is vital. Thus, establishing project objectives and other parameters such as duration, cost, risk and quality; and identifying attributes of CPS methods in terms of benefits obtained, time to schedule and technology would be needed further research to provide dynamic CPS methods selection.

3.2.3. Optimizing multi-project multi-objective schedule (MPMOS)

In previous studies, time and cost were well thought-out objectives. Researches have been examining mathematical algorithms to shorten project duration and minimize cost reduction. Ding et al. (2015) and El-Abbasy et al. (2016) developed multi-objective schedule optimization model considering multiple projects having three objectives (i.e., time, cost and quality) based on finish to start activity relationship which lacks priority weight for resource allocation. On the other hand, Elbeltagi et al. (2016) developed schedule optimization model consisting of four objectives (i.e., cost, time, cash flow and resource) for single project. Zhou et al. (2013) conducted literature based study to examine construction schedule optimization using heuristic, mathematical and metaheuristic methods. The study indicated that previous studies have focused on minimization of cost and time, and other factors such as project risk and quality have been neglected. However, organizations could perform multiple projects that are expected to meet a number of objectives such as scope, quality in dynamic environment. Since CPS are affected by various constraints such as productivity, finance, weather, available time and other constraints, further research could focus on developing multi-objective CPS models under considerable project constraints.

3.2.4. Forecasting schedule risks into activities' duration

Shen et al. (2017) carried out quantitative schedule risk analysis affecting production of prefabrication housing. The study was conducted to identify and quantitatively evaluate the interactions of major variables in CPS system and the uncertainty of each schedule risk. However, the study has the following two limitations: 1) the model did not consider perspectives of operational project management such as predecessor and successor relationship, and 2) the model established correlation among risk variables instead of duration of activities. Furthermore, Choudhry et al. (2014) identified that CPS risks highly affect project performance are in the absence of schedule mitigation guide lines. With regard to the existing schedule risk analysis methods, some are applicable in deterministic environment where there is no uncertainty; and others have been adopted to analyse the impacts of risks on project duration without any consideration of correlation between risks (Choudhry et al., 2014; Ökmen & Öztaş, 2008). More recently, there has been a little effort to predict effects of schedule risks by incorporating the correlation between activities' duration and risks. However, all of the method did not consider the correlation between schedule risks (Wang & Yuan, 2016). For this, a computerized system can be adopted with an assistance of artificial intelligence techniques such as using embedded macros to predict effects of schedule risks (Ökmen & Öztaş, 2014). Thus, further research could be put in identifying, quantifying and forecasting CPS risks by considering the correlation between schedule risks in addition to the correlation between risks and activities' duration.

3.2.5. Developing universal schedule efficiency improvement model

Evaluating the progress of CPS in comparison to baseline schedule enables to make decision on project schedule performance (Hanagodimath et al., 2016). However, many success factors affecting CPS performance such as owners' competence, conflict among stakeholders, poor human resource management, lack of knowledge and ignorance by project manager (Sinesilassie et al., 2017). Within the previous studies, forecasting and evaluating project performance were not addressed very well except few scholars. For example, Baqerin et al. (2016) carried out Weibull analysis to evaluate and forecast CPS performance analysis for repetitive construction projects. Jha and Chockalingam (2011) developed CPS performance prediction model for an ongoing construction project using few significant factors. Thus, developing universal CPS efficiency evaluation model would be promising research on developing dynamic CPS evaluation model to track, measure and forecast CPS performance for the entire project lifecycle including planning, execution, and operation.

3.2.6. Integrating emerging schedule technologies

Manual scheduling is time-consuming and prone to error despite it is one of the most integral part of construction projects (Chevallier & Russell, 1998). As a result, much research has been investigated how schedule generation could be enhanced by automating generation of activity, estimation of duration and determining sequence logic with the support of emerging technologies. For this, BIM is increasingly an important for scheduling since it provides a significant support for this complex task by reducing planning time and increase productivity (Sigalov & König, 2017). Sigalov and König (2017) developed automatic detection of construction processes patterns with an emphasis on preparation of schedules for the recognition of process patterns, including breakdown of schedules into smaller parts. Moreover, various types of spaces can be represented within 4D CAD simulation model that are required by activities in the site across their scheduled time. This enables to describe space, time and time period they exist including the volume they occupy in order to detect time-space conflict, and proactive planning prior to construction (Akinici et al., 2002). Mirzaei et al. (2018) indicated that the demonstration of labour crew movement simulation in three dimensional design geometry provides an insight for development of 4D-BIM time-space clash detection. Using the design data property such as spatial data, material layer and relationship, generation of CPS creates activities, estimate activity duration, determine sequences, and finally produce a schedule (Anderson et al., 2013). Moreover, Wang and Rezazadeh Azar (2019) developed an automated 4D-BIM CPS and Chen et al. (2011) combined BIM with genetic algorithm to assign crews for general contractor's schedule mainly for construction of building frames (i.e., wall, column, beam and slab). Vignesh (2017) has examined contractor's work flow using

LPS. Practically, there are number of trades of works that are sub-contracted for various project stakeholders such as electrical, mechanical, plumbing, heat, ventilation and other advanced building facilities to be included in BIM based CPS. Since web-based CPS are affected by various constraints such as productivity, finance, weather, available time and other constraints, further research could put in 4D-BIM and last planner to develop an integrated CPS in early stage of works and removal of constraints.

3.2.7. Application of big data analytics for examining construction schedule performance

Construction industry has been employing many scheduling tools and methods to improve construction work flows. For instance, BIM has been serving as one of the most innovative revitalization technology in the construction project schedule (Hardin & McCool, 2015). BIM data (i.e., intensive and multi-dimensional building information) is typically kept in 3D geometrical design (i.e., BIM) software products based on their specific file formats even though they use Industry Foundation Classes (IFC) as a common BIM file format. All specific file formats are opened through project management mode of particular BIM software product. Accordingly, all their files must be opened through a particular BIM software in order to visualize the three-dimensional model and various disciplines’ design property, which helps to support collabora-

tion among project stakeholders (Chang & Chen, 2015; Han & Golparvar-Fard, 2017). With an increasing recent advances in sensors and smart devices, projects thus have started generating large volume of data during the entire project life, which eventually leading to large volume of BIM data (Han & Golparvar-Fard, 2017). Due to this vast build-up of data, construction industry has recently grown beyond ‘BIM’ such as big data analytics. Bilal et al. (2016) characterized big data using three attributes: large volume of data, variety of file formats and continuous flow of data. Construction data is a diverse and voluminous due to number of BIM data, including large volume of design data and project schedule. Researchers have been dedicating to combine big data with streaming nature of construction data sources such as building information management, and creating an integrated schedule management system using GIS or BIM through the application of sensors. Thus, with the support of automatic data collection method, researchers can apply big data analytics technology to examine off-site and on-site schedule management performance.

3.3. Research trends within CPS

Based on the discussion of major research topics and gaps, the framework of future potential CPS research directions are proposed as shown in Figure 8.

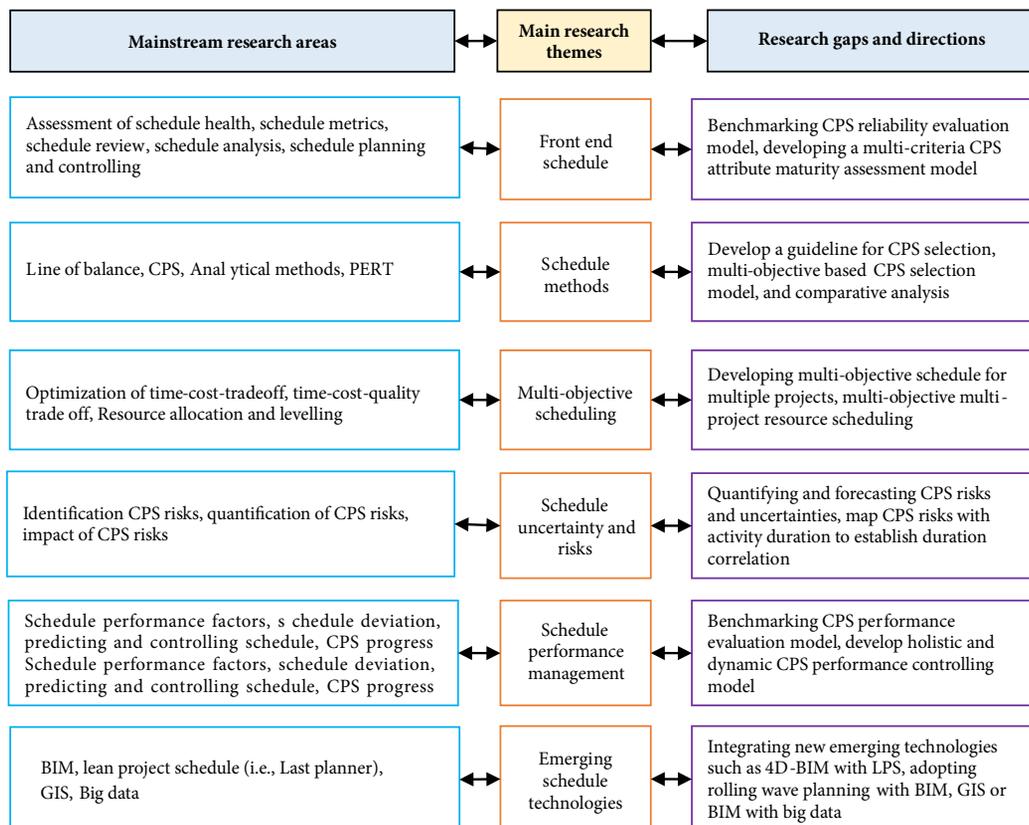


Figure 8. Comprehensive proposed framework for CPS: Research areas and directions

More promising future research directions can be summarized below:

- A comprehensive benchmarking of CPS assessment in early stage of CPS development is to validate the constructability of CPS including incorporation of required scopes represented by activity, verify that CPS risks are adequately modelled, and check logical formulation of schedule metrics;
- Forecasting robust CPS still needs research in development of front-end CPS reliability model together with multidimensional aspects of project such as scope, risks, schedule metrics and reasonable estimation of duration prior to the start of construction to detect the deficiency of CPS;
- Investigating on taxonomy of CPS methods by Al Nasser et al. (2016) has been conducted. It has been identified that further research is required in developing an optimized CPS method selection model using identified parameters usability, extent of adoptability, limitations and coverage of each methods in relation to project scope such as size and complexity;
- Studies in achieving multiple objectives for multiple projects at the same time could drive motivation for further research. For example, multi-objective schedule optimization model for multiple projects (El-Abbas et al., 2016), multi-objective schedule optimization for single project (Elbeltagi et al., 2016). It is anticipated that these previous studies information plus the application of modern optimization algorithms would be more developed in future research of CPS for multiple projects' objective within unknown project uncertainties;
- A more comprehensive quantification system of CPS risks needs to be developed to accurately predict the probable effects of CPS risks. Even though, identifying an extent of impacts of schedule risks provides the probable effects of schedule risks on overall project duration, mapping CPS risks into activity duration is more binding to identify most sensitive activities. Thus, forecasting CPS risks based on impacted activity duration could put future research in the domain of CPS using proposed framework as shown in the Figure 8, and quantification of CPS risks initiated by Liu et al. (2015);
- An advanced CPS planning and management system can be incorporated for the entire life cycle of project starting in the early stage of project planning and design stage. Developing 4D-BIM-Last Planner system using an integration of 3D-design geometry and rolling wave planning principle would be implemented in order to reduce CPS risks and uncertainties;
- Examining off-site and on-site schedule management performance using Big data analytics technology with the support of automatic construction data collection systems such as sensors and smart devices.

Conclusions

In this paper, holistic literature review-based study was implemented in the domain of construction project schedule. A bibliometric literature retrieval and scientometric analysis were used with detailed qualitative explanation. A total of 332 journal articles were chosen since 2009. The scientometric analysis illustrates the most prominent journals, keywords, co-author analysis, articles, and countries/regions in the research of CPS.

Based on scientometric analysis followed by detailed qualitative discussion, main research topics and research gaps in the domain of CPS are identified; and simultaneously future research directions are proposed. The major research topics include mathematical algorithms-based resource scheduling optimizations, CPS methods and applications, identification and quantification of CPS risks, and CPS planning and controlling. The gaps of existing research studies include front-end CPS constructability assessment, optimizing CPS methods selection, multi-objective for multiple project schedule optimization, mapping and forecasting of CPS risks into activity duration, CPS performance evaluation, and integrating emerging schedule technologies. The study also proposes future research directions.

This review-based study in CPS was limited to bibliometric literature samples. The study only focused journal articles written in the English language in the domain of CPS academic research areas in Scopus database source. Other sources, including conferences were excluded. Thus, there would be further research to discover the uncertainty between the latest industry practices and the scholarly research.

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Author contributions

Mr. Derbe was responsible for data collection from Scopus database, results analysis, interpretation, discussion and conclusion. Dr. Li conducted the detailed construction project schedule (CPS) analysis and wrote the first draft of the article. Dr. Wu checked the related articles from Google Scholar, Scopus and Web of Science and provided the corresponding analysis. Dr. Zhao proposed the review method scheme and refined the submission to Journal of Civil Engineering and Management.

Disclosure statement

We affirm that we have no any financial, personal, or professional interests from other parties.

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