Digital twins: artificial intelligence and the IoT cyber-physical systems in Industry 4.0

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Abstract
This paper presents a summary of mechanisms for the evolution of artificial intelligence in ‘internet of things’ networks. Firstly, the paper investigates how the use of new technologies in industrial systems improves organisational resilience supporting both a technical and human level. Secondly, the paper reports empirical results that correlate academic literature with Industry 4.0 interdependencies between edge components to both external and internal services and systems. The novelty of the paper is a new approach for creating a virtual representation operating as a real-time digital counterpart of a physical object or process (i.e., digital twin) outlined in a conceptual diagram. The methodology applied in this paper resembled a grounded theory analysis of complex interconnected and coupled systems. By connecting the human–computer interactions in different information knowledge management systems, this paper presents a summary of mechanisms for the evolution of artificial intelligence in internet of things networks.

Keywords Digital twin · Industrial internet of things · Cyber physical systems · Human and robot interactions · Industry 4.0 · Bibliometric analysis

1 Introduction
The volume of data generated at the edge, creates diverse challenges for developing data strategies in a variety of areas (e.g., artificial intelligence, ethics, edge vs fog computing). Data strategies for complex coupled systems at the edge, demand data strategy optimisation for collecting of probabilistic data, when edge computing nodes are deployed. The personal perceptions of risk as a result of collecting probabilistic data at the edge interact with data regulations, standards and policies. The focus of the empirical research (in Sect. 3) of this study is the integration of these data perceptions, regulations and policies for conceptualising the creation of ‘digital twins’. Our research objective is to apply statistical methods on existing research records and to identify how artificial intelligence (AI), is evolving at the edge. With consideration of the above, our research focus was on AI integration in ‘cyber-psychnical systems’ (CPS) that connect human–computer interactions. Our research emphasis was on conceptualisation of AI assimilation in ‘digital twins’ of CPS—that meets public acceptability, security standards, and legal scrutiny. This paper builds upon earlier literature on cyber risk in Industry 4.0 (I4.0) (Radanliev et al. 2020a, b, c, d), but with less focus on supply chains and much greater emphasis on statistical biometrical analysis of large number of data records and qualitative empirical analysis of the leading Industry 4.0 national frameworks.

To address these research objectives, the paper integrates the topics of artificial intelligence and cyber physical systems with areas including ‘Industry 4.0’ (I4.0)—which represent the digitalisation of industrial processes; ‘internet-of-things’ (IoT)—which is defined as connected system (Fig. 1).

The paper builds upon established methodologies (1) literature review and bibliometric analysis to identify concepts from existing research. (2) empirical review of most
documented frameworks and initiatives on Industry 4.0; (3) grounded theory, to identify and group concepts from the literature review, bibliometric analysis, and the empirical analysis, followed by (4) conceptual diagram design to present the new process.

The reminder of the paper is structured as follow: In Sect. 2 we present a literature review on challenges for privacy, security and resilience of connected environments, and future challenges for AI systems embedded within IoT. We conclude Sect. 2 with a bibliometric analysis, using statistical software and R studio to present three-fields plot, collaboration network map, and conceptual research structure map with using factorial analysis. In Sect. 3 we present our empirical analysis, where the results of applying grounded theory, for conceptual groupings, are presented in summary tables and followed by conceptual diagram illustration. Followed by a discussion in Sect. 4, and conclusions in Sect. 5.

2 Literature review

The objective of the qualitative investigation is to review a selected article that are considered as leading academic research in this area. The objective of the quantitative analysis is to review all current and historical data records. In our literature review and bibliometric analysis, we followed recommendations on ensuring ‘relevance’ of the analysed literature (Lezzi et al. 2018), accuracy through ‘text mining’ (Rivas et al. 2018), with a focus on ‘integration’ of different technologies (Sittón-Candanedo 2020).

2.1 Present challenges for privacy, security and resilience of connected environments

Present challenges for artificial intelligence in industrial techniques are in the areas of: robustness, safety, and security (Hahn et al. 2013; Zhu et al. 2011); control and hybrid systems (Leitão et al. 2016; Shi et al. 2011); computational abstractions architectures (Madakam et al. 2015); real-time embedded systems abstractions (Marwedel and Engel 2016; Radanliev et al. 2020a); model-based development (Jensen et al. 2011) and education and training (Faller and Feldmüller 2015). However, as the artificial cognition in cyber-physical systems (CPS) is an evolutionary process (Wahlster et al. 2013), the challenges will be evolving with time (Radanliev et al. 2020a, b, c, d), which requires flexibility management of the complexities of the CPS. In addition to the evolving challenges, for creating a reliable, secure and economically sustainable power system, financial planning arrangements should be developed for CPS applications, e.g. Sentinel (Balaji et al. 2015)—can be used to exploit the information flow for energy savings.

2.2 Future challenges for AI systems embedded within IoT

Building upon the present challenges, futuristic artificial cognition in CPS requires the deployment of self-sustaining networked sensors and actuators (Rajkumar et al. 2010) operating in symbiotic relations with the physical environment modelled through a user-centric Cloud (Gubbi et al. 2013). Environmental natural resources are crucial in sustaining economic development (Stock and Seliger 2016), and CPS in the future should be focused on creating eco-industrial by-product synergy (Pan et al. 2015).

Another challenge is the evolution into self-adapting cognitive decentralised CPS. This process would present a distributed (Wan et al. 2015) and integrated-decentralised CPS swarm intelligence, based on cooperation of large population of simple agents (Stojmenovic 2014). The decentralisation of the control function to a swarm—which implies lack of central control dictating individual behaviour and interactions among swarm agents—stimulates the emergence of intelligent global behaviour. Individual agents exploit cloud services to analyse, predict, optimise and mine scalable capabilities of historical data and enable applications to self-adjust their behaviour to self-optimise their own performance (Brettel et al. 2016; Shafiq et al. 2015). In such decentralised systems, individual agents’ ‘contract-based design’ is applied before ‘platform-based design’ (Sangiovanni-Vincentelli et al. 2012).
Contract-based design enables actor-oriented design of multiple models of computation to be integrated in a single hierarchical system (Bhave et al. 2011), like loosely time-triggered CPS architectures (Benveniste 2010).

3 Methodology for the bibliometric analysis of existing literature with computable statistical methods

In this section, we analyse existing literature records with computable statistical methods. While the literature review was focused on qualitative review of selected articles, this chapter analyses data records with statistical categorisations that eliminate bias of qualitative interpretations, and data records selection. To analyse the present and future challenges identified in the literature review, we conducted a bibliometric search for data records. We searched the Web of Science Core Collection for records on the TOPIC: (cyber physical systems) AND TOPIC: (internet of things) AND TOPIC: (industry 4.0). This search resulted with 436 in a timespan from 1900 to 2020. We obtained the data from the Web of Science with a Boolean search that include only ‘AND’. In the spirit of reproducible research, we note that if similar search is conducted, but instead of AND, we use OR, then the results would be 60,148. These records would however not be related to a research related to all the topics of IoT, CPS and I4.0, but instead, the data records would be relevant to individual topics. Since in this study, we wanted to investigate the relationship between these systems, we used the smaller sample of 436 data record. This was based on our rationale to investigate the relationship between these topics, and not the research on these topics in isolation. Since we wanted to investigate the tacit evolution of AI in these connected systems, we did not include AI in the data records search, instead, our investigation was on identifying if concepts related to AI are present in the data records. We wanted to investigate if AI is evolving naturally as a result of the relationships between these connected systems. To analyse the data records, we used statistical methods with R studio and the ‘bibliometrix’ package (Aria and Cuccurullo 2017), see Fig. 2.

In Fig. 2, we designed a three-fields plot of the 436 data records, and we can see the connections between different research topics by keywords, by fields, and by countries. In the spirit of reproducible research and eliminating bias in research, we include one figure analysis where the data search and analysis is conducted with the Web of Science research analysis tool, but instead of AND, we use OR, and we analyse the results of 60,148 records on these subjects, but with individual topic investigated, which can be replicated by searching for the TOPIC: (cyber physical systems) OR TOPIC: (internet of things) OR TOPIC: (industry 4.0), in Fig. 3.

In Fig. 3, we can see that China has produced the most scientific research on the individual topics in isolation.
While on the three-field plot (Fig. 2) we can see that US and China have produced a comparative number of records on these topics in combination, what is puzzling is that in Fig. 2, the UK and Brazil are also strongly represented. While in Fig. 3, Germany and Italy score higher than the UK, and Brazil is not even in the top countries selected by the We of Science research analysis of the 60,148 records. Therefore, we continued the statistical analysis of the bibliometric records. Although the three-fields plot in Fig. 2 is very detailed in visualising the connections between research keywords, fields, and relating them by countries, we could not identify the collaborations between countries. To analyse which counties collaborate in the fields of IoT, CPS, and I4.0, we created a collaboration network map Fig. 4.

From the collaboration network map, we can see the historical and present collaboration relationships on the topics of IoT, CPS and I4.0. For example, China is working closely with Singapore, a cluster is visible between Brazil, Portugal, Spain, France and Romania, and a second cluster between Germany, Turkey, Sweden and Austria. These clusters are expected, because they also represent known clusters based on culture, economy, tourism, etc. However, the most unexpected cluster is visible in the red colour, strongly representing the UK, USA and China. Since the UK has been in the EU until 2019, one would have expected a stronger cluster between the UK and other EU countries, like Germany. Perhaps this can be explained by language and cultural similarities between the UK and USA. The relationship between the US and China, could possibly be explained by economic factors. To analyse further this data records, we developed a conceptual structure map in Fig. 5.

The conceptual structure map in Fig. 5 is designed with applying factorial analysis. We wanted to see how closely are related the keywords and research fields from the three-fields plot in Fig. 2. In the three-fields plot, we could only see the connections, but we could not see how closely if these areas are related. In the Fig. 5, conceptual structure map, we can see how close or distant are specific research areas. We complimented the statistical analysis with literature review analysis to categorise emerging themes for embedding AI systems within IoT systems.
4 Categorisation of future and present techniques: found in the qualitative literature review and the quantitative bibliometric analysis

This review presents different groups of future challenges and techniques for embedding AI systems within IoT systems, and a form of dependency or a causal edge relation between them. Table 1 shows the edges between different nodes representing groups of the influence of future and present challenges and techniques on artificial cognition in CPS.

The techniques identified (in Table 1) represent a form of conceptual grouping for building upon the CPS architecture (Lee et al. 2015), but separates the future from the present techniques and challenges. This separation constitutes the grounding for the empirical review of AI systems within IoT systems summarised in Tables 2, 3 and 4, presenting an overview of I4.0 strategies in tabular form. Table 1 is also important for the understanding of the conceptual diagram, in Fig. 6, because all abbreviations are explained in Table 1.

5 Empirical review: embedding IoT systems in I4.0 frameworks

The aim of this section is to relate the academic literature and industry reports, with what is happening in practice. The first objective of the empirical review is to determine whether the practical frameworks and initiative globally are implementing the recommendations from the leading research in this area, as identified in the literature review—Sect. 2. The second objective is to relate the findings from the bibliometric analysis, with the leading frameworks and initiatives, and to build summary tables of the emerging categories. The summary tables are used for the conceptual diagram design, illustrating the evolution of Industry 4.0 and IoT enhanced cyber-physical systems. Finally, by empirically comparing academic literature with national frameworks, we propose a practical approach, which is considered crucial in current literature (Radanliev, De Roure, Nurse, et al., 2020), and also discussed in earlier supply chain literature (Radanliev 2016).
This diversity advances to the summary tables of the less evolved world initiatives grows in magnitude as the empirical analysis. The diversity of the approaches in architecture, with a step-by-step method for integration of CPS some of the complexities in developing a unifying architecture and follows with an empirical review of (additional 13) I4.0 emerging from the empirical review, capture different and often conflicting approaches that harden adoption of coherent standards for artificial cognition in CPS.

The conceptual diagram in Fig. 6 separates the future challenges from the present challenges in the three summary tables (Tables 2, 3, 4). The categorisations in the summary tables (Tables 2, 3, 4), detail our grounded theory process, where emerging concepts are extracted from the large number of digitalisation initiatives from around the globe. The emerging concepts are structured in a conceptual diagram, using categorical coding in Fig. 6. To understand better the conceptual diagram, in Fig. 6, we refer to the abbreviations in Table 1.

In Fig. 6 to simplify the differences between areas of: ‘focus’, ‘decision’ and ‘action’, we used the world leading initiatives, and grouped concepts based on similarities, e.g. in the key technologies: NIF lists 47 key technologies, Made in China 2025 prioritises 10 tech sectors and NTI includes 13 key tech examples. However, some initiatives focus on areas of decision and action that differ greatly from the main objectives of other initiatives. For example, the focus on market networks and market creation of the National Technology Initiative (ASI 2016) is opposed to the mainstream technology development in NIF that includes policies for subsidies and repayable advances, tax incentives and loans. These differences emphasised the need to develop the summary tables that serve as a mechanism for inter-relating the three areas of focus, decision and action in a meaningful method. The analysis outlined in Tables 2, 3 and 4 provides detailed explanation of these areas.

6 Discussion

Through empirical and statistical analysis, we presented a new understanding on how artificial intelligence technologies are migrated to the periphery of the internet and into local internet of things networks. With qualitative

| Table 1 Future and present techniques for embedding AI systems within IoT systems |
|---------------------------------|---------------------------------|
| Present techniques and challenges |
| Education and training | EaT |
| Financial planning | FP |
| Information flow for energy savings | IFIES |
| Robustness, safety, and security | RSS |
| Control and hybrid systems | CHS |
| Computational abstractions architecture | CAA |
| Real-time embedded systems abstractions | RTESA |
| Model-based development | MBD |
| Future techniques and challenges |
| Eco-industrial by-product synergy | EIS |
| Distributed integrated-decentralised | DID |
| Multi-agent swarm intelligence | SI |
| Contract-based design | CBD |
| Self-sustaining networked sensors | SSS |
| Symbiotic relations with the physical environment | SRPE |
| User-centric Cloud based vision | CCV |
| Future vision for artificial cognition in CPS | EABaC |
| Energy-aware buildings and cities | CIPM |
| Critical infrastructure with preventive maintenance | SCCPS |
| Self-correcting of cyber-physical systems | FACTS |
| Flexible AC transmission systems | DER |
| Distributed energy resource technologies | SELF |

The empirical review starts with the German I4.0 (Industrie 4.0) (Germany Trade Invest 2014; Industrie 4.0, 2017) and follows with an empirical review of (additional 13) I4.0 initiatives. The main elements of each initiative in Table 3 are separated in: areas of focus, decision and action. The empirical review identifies several shortcomings in individual initiatives, which are complimented by other initiatives. This required building a model that integrates the strengths and reduces the weaknesses of all initiatives. There are problems, when, for example, some of the areas of focus, decision and action (Tables 2, 3, 4) differ in terms of strategy and propose very different approaches. To resolve this issue, we use two strategies. Firstly, the individual areas as categorised in the empirical summary tables that are used as reference categories. Secondly, the summary tables are used for embedding AI systems within IoT systems in I4.0 mechanism that relates various areas to each other and eliminates conflicts in different and sometimes contrasting approaches.

The summary table analysis of the initiatives presents some of the complexities in developing a unifying architecture, with a step-by-step method for integration of CPS cognitive capabilities. The diversity of the approaches in world initiatives grows in magnitude as the empirical analysis advances to the summary tables of the less evolved (Table 3) and the elusive initiatives (Table 4). This diversity is categorised with the grounded theory, into areas of ‘focus’, ‘decision’, and ‘action’. Some initiatives contain concepts that are based on prioritising, and we created a category called ‘priority’ for such concepts. Or prioritised ‘relationships’, so we created a separate group for such concepts. In the summer tables, we present a multiverse of such categories that emerge from applying the grounded theory in the empirical analysis. Some of these areas appear unrelated, and yet, they are present. Therefore, the relationship is not clearly presented in the national initiative, but it becomes clearer in the summary table, when placed in a category against other similar concepts.

The differences in the categorised approaches in the three summary tables correspond to the different national and international strengths and aspirations of the promoters of these strategies. The three summary tables (Tables 2, 3, 4) emerging from the empirical review, capture different and often conflicting approaches that harden adoption of coherent standards for artificial cognition in CPS.
| 14.0 national frameworks | Category 1: Areas of focus for artificial cognition in CPS | Category 2: Areas of decision for artificial cognition in CPS | Category 3: Areas of action for artificial cognition in CPS |
|-------------------------|----------------------------------------------------------|----------------------------------------------------------|----------------------------------------------------------|
| Germany—Industrie 4.0 (Germany Trade Invest 2014) | **Workgroups—14.0 policy:** (1) The Smart Factory; (2) The Real Environment; (3) The Economic Environment; (4) Human Beings and Work; (5) The Technology Factor | **Relationships—14.0 principles:** (1) based on CPS, IoT and Cloud Computing | **Priority areas—14.0 mission/vision—the Smart Service 5 priority areas:** (1) Integrated Production and Service Innovation; (2) Internet and Service Economy; (3) Technological Enablers; (4) Business Organisation Requirements; (5) Innovation-oriented Framework |
| | | 5C imperative element—6C system: (1) Connection; (2) Cloud; (3) Cyber; (4) Content/context; (5) Community; (6) Customisation | 5C key testbeds for 14.0 technologies—Industrie 4.0 research into practice, over 500 projects are being carried out in Germany (Industrie 4.0, 2017) |
| | | 5C design building blocks for 14.0—4 key design principles: (1) Interoperability; (2) Information transparency; (3) Technical assistance; (4) Decentralised decisions | |
| USA (1) Industrial Internet Consortium (Industrial Internet Consortium 2017) (2) Advanced Manufacturing Partnership (NIST Advanced Manufacturing Office 2013) | **5C key industries for 14.0—IIC industries:** (1) Energy; (2) Healthcare; (3) Manufacturing; (4) Smart Cities; (5) Transportation | | **5C key testbeds for 14.0 technologies—IIC testbeds:** 22 Testbeds to be deployed for the Industrial Internet; |
| | | | | **Priority areas—AMP snapshot of priority technology areas—AMP mission/vision:** (1) Advanced materials manufacturing; (2) Engineering biology to advance bio-manufacturing; (3) Bio-manufacturing for regenerative medicine; (4) Advanced bio-products manufacturing; (5) Continuous manufacturing of pharmaceuticals |
| | | | | **5C imperative element—AMP areas of interest:** (1) Advanced Sensors, Controls, Platforms, and Modelling (ASCPM); (2) Visualisation, Informatics, and Digital Manufacturing Technologies (VIDM) (Edgar and Davis, 2015) |
| | | | | **5C design actions for 14.0—AMP goals:** (1) Increase the competitiveness of U.S. manufacturing; (2) Facilitate the transition of innovative technologies into manufacturing; (3) Accelerate the development of an advanced manufacturing workforce; (4) Support sustainable business models |
| | | | | **5C design building blocks for 14.0—AMP areas of focus:** (1) Technology readiness; (2) UK policy; (3) Funding from 3 sources: public funding, business contracts, collaborative projects |
| | | | | **5C key projects for 14.0—Manufacturing catapult 7 Key projects:** 7 HVM Catapult centres (John, 2017) |
| | | | | Digital Strategy: (1) Connectivity; (2) Digital skills; (3) Digital sectors; (4) Wider economy; (5) Cyber security; (6) Digital government; (7) Data economy |
| UK—(1) Catapults (John 2017); (2) UK Digital Strategy (Department for Digital Culture Media Sport 2017) | **Priority area—DC mission/vision: Catalyst for sustainable high value manufacturing** | | **5C key testbeds for 14.0 technologies—cata- pacuts:** (1) 11 Digital Catapult centres—including HVM network of another seven centres, (5C key technologies for 14.0) (2) 27 Key technologies |
| | | Measured area of impact: Economic impact | **5C key technologies for 14.0:** (2) 27 Key technologies |
| | | 5C design actions for 14.0—areas of focus: (1) Technology readiness; (2) UK policy; (3) Funding from 3 sources: public funding, business contracts, collaborative projects | | **5C key projects for 14.0—Manufacturing catapult 7 Key projects:** 7 HVM Catapult centres (John, 2017) |
| | | | | Digital Strategy: (1) Connectivity; (2) Digital skills; (3) Digital sectors; (4) Wider economy; (5) Cyber security; (6) Digital government; (7) Data economy |
| | | | | **5C design building blocks for 14.0—design elements:** (1) Open access facilities; (2) Contractual agreements; (3) Intellectual property and confidentiality; (4) State aid compliance; (5) Financial support; (6) Publicity |
| | | | | **I4.0 focus—principles:** (1) Survey; (2) Industrial strategy consultation |
| | | | | Digital Strategy: (1) Connectivity; (2) Digital skills; (3) Digital sectors; (4) Wider economy; (5) Cyber security; (6) Digital government; (7) Data economy |
| 14.0 national frameworks | Category 1: Areas of focus for artificial cognition in CPS | Category 2: Areas of decision for artificial cognition in CPS | Category 3: Areas of action for artificial cognition in CPS |
|--------------------------|----------------------------------------------------------|----------------------------------------------------------|----------------------------------------------------------|
| Japan—(1) Industrial Value Chain Initiative (Industrial Value Chain Initiative Japan 2017) (2) New Robot Strategy (NRS) (Ministry of Economy Trade and Industry of Japan 2015) and RRI (Ministry of Economy 2015) | **5C design building blocks for I4.0—IVI**
loosely defined standards: (1) Industrial Value Chain Reference Architecture (IVRA); (2) IVI Platform for Smart Manufacturing Ecosystem (IVI, 2017); (3) Smart Manufacturing Business Scenarios (IVI, 2016)
**NRS pillars**—5C imperative element: (1) Robots innovation hub; (2) Robot society; (3) Robotics in IoT
**NRS areas of focus:** (1) AI learning from big-data; (2) AI reasoning from existing knowledge; (3) AI brain | **Relationship**—IVI strengths/principles: (1) Gather a broader understanding of more general connection models (reference models); (2) Aims to design ‘loosely defined standard’, as an adaptable model instead of a rigid system
**5C design actions for I4.0—NRS objectives:** (1) AI reasoning; (2) AI learning; (3) AI intelligence infrastructure | **IVI Activities—policy:** (1) IVI Forum; (2) 25 Business Scenario Workgroups (BSWGs); (3) 8 Platform Workgroups (PWGs)
**Priority areas**—NRS mission/vision: (1) Robots autonomy; (2) Robot as information terminals; (3) Robotic networking
**5C key technologies for I4.0—NRS core technologies:** (1) Artificial intelligence; (2) Automated behaviour |
| France—New France Industrial (NFI)—also known as: la Nouvelle France Industrielle or Industry of the Future (New Industrial France 2016) | **5C key markets for I4.0—key Markets**—9 solutions: (1) Data economy; (2) Smart objects; (3) Digital trust; (4) Smart food production; (5) New resources; (6) Sustainable cities; (7) Eco-mobility; (8) Medicine of the future; (9) Transport of tomorrow | **NIF policy:** (1) Subsidies or repayable advances; (2) Tax incentives; (3) Loans; (4) SME business modelling support
**Priority areas**—Pillars—NIF mission/vision: (1) Developing cutting edge technologies; (2) Helping companies adapt to the new paradigm; (3) Training employees; (4) Showcasing the French industry of the future; (5) Strengthening European and international cooperation | **5C key technologies for I4.0—key technologies:** List of 47 key technologies that need to be industrialised in the 9 solutions
**5C imperative element:** NFI economy of data: (1) Digital technology, virtualisation and the Internet of Things; (2) Cobotics, augmented reality; (3) Additive manufacturing (3D printing); (4) Monitoring and control; (5) Composites, new materials and assembly; (6) Automation and robotics; (7) Energy efficiency (NIF, 2013) |
investigation, we identified that artificial intelligence improves organisational resilience in industrial systems, supporting both a technical and human level. With empirical analysis, we correlated the world leading Industry 4.0 frameworks, and modelled the interdependencies between edge components to both external and internal services and systems in summary tables. The methodology we applied resembled a grounded theory analysis of complex interconnected and coupled systems. This enables us to connect the human–computer interactions in different information knowledge management systems with artificial intelligence, and to design a mechanism for future evolution of artificial intelligence in internet of things networks. We argue that such evolution is inevitable, and already happening in Industry 4.0 automations.

In this review study, we combined qualitative review of recent and relevant literature from Google Scholar, with quantitative computable statistical analysis of bibliometric data records from the Web of Science, and empirical review of government frameworks. The findings from this review study is an outline of present and future research challenges in Sects. 2.1 and 2.2, analysed with the bibliomic analysis in Sect. 3 and summarised in Table 1. In Table 1, we can see the present and future research challenges categorised and abbreviated as terminology for building the conceptual diagram in Fig. 6, presenting the state-of-the-art in our understanding on how AI evolves within the IoT.

The implications for research are a conceptual approach that enables us to connect the human–computer, with the evolution of artificial intelligence in connected networks. We argue that such evolution is inevitable, and already happening in Industry 4.0 automations.

The future research dimensions emerging from this study are related to enhancing our capacity for understanding the opportunities when AI technologies are integrated in smart connected networks. To assist in these future dimensions, the summary tables developed throughout the paper, can serve as a best practice and inform initial steps taken for design and prototype of AI-enabled IoT. AI-enabled IoT would enable a comprehensive and systematic understanding of the opportunities and threats that arise when AI technologies are migrated to the periphery of the internet and into local IoT networks. The summary tables detail a new CPS process, that enables practitioners to understand and design deep learning solutions for embedding AI systems. Recent research has discussed the integration and the cyber risks from Fog computing (FC) and Internet of Everything (IoE) (Radanliev et al. 2020a). This research was focused on CPS process in Edge computing and Internet of Things. The new CPS process can be applied in real systems like smart cites and 5G technology, with the integration of IoT connected systems. These systems (IoT and CPS) are interconnected and the results in Fig. 1 present hierarchical categories integrating the two systems.

The main implications for practitioners from this study, is the new CPS process. When enhanced with recent deep learning solutions, the new CPS process can be applied in real systems like smart cites and 5G technology. The new CPS process can also help governments trying to improve and/or build their national strategies. The value from the developed mechanisms for national governments of the developed nations, is that they can check and compare the existing digital strategies with the national digital strategies applied across the globe.

The findings from this study benefit digital industries in developed and developing nations, with a design process describing how to use the new CPS process in industrial applications. The benefits for digital industries in developing countries that have no digital strategies (e.g., most African, Latin American, Eastern European nations), is that they can review the current developments from around the world, and follow the step by step process to embed AI systems within IoT systems of their I4.0 frameworks. The benefits for digital industries in developed countries is included in the summary tables, which are presented in this paper and are designed to support the building of future artificial cognition in CPS and the improving and reformulating of existing I4.0 frameworks and practical initiatives. The summary table mechanism emerging from the empirical review is based on grouping of future and present techniques and presenting the design process for embedding AI systems within IoT systems. These are established models for decomposing and
| 14.0 national frameworks                      | Category 1: Areas of focus                                                                 | Category 2: Areas of decision                                                                 | Category 3: Areas of action                                                                 |
|----------------------------------------------|-------------------------------------------------------------------------------------------|---------------------------------------------------------------------------------------------|---------------------------------------------------------------------------------------------|
| Nederland—Smart Industry; or Factories of the Future 4.0 (Bouws et al. 2015) | **Priority areas**—mission/vision: (1) Defining strategic objectives; (2) Defining activities within the smart industry agenda; (3) Defining implementation | **Relationship**—principles: (1) SWAT analysis; (2) Developed a ‘dare to share’ data cooperation initiative | **5C design building blocks for 14.0**—key contribution: (1) Demonstrates illustrative projects; (2) Demonstrates data sharing initiatives; (3) Developed action objectives, agenda and plan on standardisation; (4) Developed action line on cyber security |
| Belgium—Made Different (Sirris and Agoria 2017) | **Priority areas**—Human centred dynamic production—mission/vision: (1) High value market responsive manufacturing; (2) New business models and digitised production; (3) On demand resilient production system; (4) CPS; (5) Circular economy; (6) Reduce materials and energy consumption | **Relationship**—principles: (1) Based on CPS; (2) blog (forum); (3) Focused on products and services with high added value | **5C design building blocks for 14.0**—promotes 7 crucial transformations: (1) World Class Manufacturing Technologies; (2) End-to-end Engineering; (3) Digital Factory; (4) Human Centred Production; (5) Production Network; (6) Eco Production; (7) Smart Production Systems |
| Spain—Industrie Conectada 4.0 (Ministry of Economy Industry and Competitiveness Accessibility, 2015) | **Priority area**—area of focus: mission/vision Linking the physical to the virtual to create intelligent technology | **Principles**: (1) Based on CPS; Policy: (2) Financial support for digital transformation; (3) Personalised advice for SMES; Relationship: HADA—Advanced Self-diagnosis tool | **5C design building blocks for 14.0**—promotes: (1) Hybridisation between the physical and digital worlds; (2) Digital transformation for the evolution to the digital economy |
| Italy—Fabbrica Intelligente (Ministry of Education Universities and Research 2014) | **Priority areas**—areas of focus—mission/vision: Developing and implementing a strategy for (1) Transforming towards new product, services, process, and technologies; (2) Design, execution and enhancements of research results; (3) Connecting nation and regional research with international policies; (4) Improving possibilities for using EU funds | **5C design building blocks for 14.0**—design activities: (1) Realisation of research projects; (2) Technology transfer, sharing of knowledge and networking; (3) Sharing research infrastructure; (4) Support and facilitation of smart and sustainable entrepreneurship; (5) Support for technological forecasting activities in the smart factory sector; (6) Support for the growth of the human capital | **5C key projects for 14.0**—4 Key projects: (1) Sustainable manufacturing; (2) Adaptive Manufacturing; (3) Smart Manufacturing 2020; (4) High Performance Manufacturing |
Table 4  Summary table from the empirical analysis of roughly defined I4.0 mechanisms for embedding AI systems within IoT systems

| I4.0 national frameworks                                      | Category 1: Areas of focus                                                                 | Category 2: Areas of decision                                                                 | Category 3: Areas of action                                                                 |
|---------------------------------------------------------------|-------------------------------------------------------------------------------------------|---------------------------------------------------------------------------------------------|-------------------------------------------------------------------------------------------|
| China—Made in China 2025 (The State Council People Republic of China 2017) | **Priority area**—mission: Comprehensive upgrade of the Chinese industry                | **Role of the state**—policy: (1) Provide an overall framework, (2) Utilising financial      | **5C key tech sectors for I4.0—10 priority tech sectors:** (1) New advanced information    |
|                                                               | Guiding principles—mission/vision: (1) Innovation-driven manufacturing; (2) Quality over  | and fiscal tools, and supporting the creation of manufacturing innovation centres (15 by     | technology; (2) Automated machine tools & robotics; (3) Aerospace and aeronautical-        |
|                                                               | quantity; (3) Green development; (4) Optimise the Chinese industry infrastructure; (5)    | 2020 and 40 by 2025)                                                                        | equipment; (4) Maritime equipment and high-tech shipping; (5) Modern rail transport         |
|                                                               | Nurture human talent                                                                      | **5C design building blocks for I4.0—role of the industry:** (1) Strengthening intellectual  | equipment; (6) New-energy vehicles and equipment; (7) Power equipment; (8) Agricultural      |
|                                                               | **Difference—principles:** (1) Focus on the entire manufacturing process and not just     | property rights; (2) Protection for small and medium-sized enterprises (SMEs); (3) Allowing  | equipment; (9) New materials; and (10) Biopharma and advanced medical products            |
|                                                               | innovation; (2) Promotes the development of advanced and traditional industries and        | companies to self-declare their own technology standards; (4) Help companies participate    |
|                                                               | modern services; (3) Less focus on state involvement, and market mechanisms are more      | in international standards setting                                                            |
|                                                               | prominent; (4) Clear and specific measures for innovation, quality, intelligent          |                                                                                             |
|                                                               | manufacturing, and green production                                                      |                                                                                             |
| G20—New Industrial Revolution (NIR) (G20 2016)               | **Priority area**—mission/vision: Promoting joint action by enhancing existing           | **Policy key point**—policy: (1) Trade liberalisation and the elimination of subsidies       | **5C design actions for I4.0—actions:** (1) Research collaboration; (2) The role of SMEs; |
|                                                               | communication and collaboration mechanisms                                                |                                                                                             | (3) Employment and workforce skills; (4) Cooperation on standards; (5) New industrial       |
|                                                               | NIR principles: develop multi-stakeholder communication principles within and across      |                                                                                             | infrastructure; (6) Intellectual property rights protection; (7) Industrialisation in       |
|                                                               | countries                                                                                |                                                                                             | developing countries                                                                       |
| Russia—National Technology Initiative (NTI) (ASI 2016)        | **Priority areas**—mission/vision: (1) Identifying new markets; (2) Identifying key      | **Policy key point**—policy: Promote market demand, new technological standards and national | **5C key tech networks for I4.0—key technologies:** 13 key tech examples                    |
|                                                               | technologies                                                                             | cyber security                                                                                 | **5C key tech networks for I4.0—set of market network:** (1) AutoNet; (2) AeroNet; (3)   |
|                                                               | Guiding key principle: Focus is on market creation as opposed to technology development  | 5C design building blocks for I4.0—strength: Electronic open submission of recommenda-     | MariNet—national security resources; (4) SafeNet; (5) EnergyNet; (6) FoodNet—technological  |
|                                                               |                                                                                         | tions for changing or editing the strategy                                                  | changes; (7) HealthNet; (8) NeuroNet; and (9) FinNet                                    |
reverse engineering design processes that are present in this paper.

Validity of the research findings was pursued through applying the grounded theory approach (Goulding 2002), to create the categorisations in the summary tables, which is a time tested method for building theory from existing knowledge. For ensuring validity of the findings, we also applied qualitative research techniques, including open and categorical coding (Easterby-Smith et al. 2002; Eriksson and Kovalainen 2008; Gummesson 2000), in the summary tables. Open coding to validate a reliable representation of the data records, categorical coding to recognise the concepts in the data (Goulding 2002). Discourse analysis is applied to evaluate and interpret the connotation behind the explicitly stated approaches (Eriksson and Kovalainen 2008), along with tables of evidence (Eisenhardt 1989) as seen in the summary tables, and conceptual diagram (Miles et al. 1983) for graphical analysis.

7 Conclusion

The research integrates concepts related to digital automation (e.g., digital twins), that have not been previously integrated in a research. Through statistical analysis of bibliometric data records from 1900 to 2020, this study identified how artificial intelligence, is evolving in internet of things systems. With factorial analysis, we identified the main national clusters of research, on how artificial intelligence, is evolving in internet of things systems and cyber-physical systems. From our analysis, it becomes clear that despite political turbulences, the US is closely coupled in the research collaboration network with China. Furthermore, despite political and economic collaborations, that even include a joint research budget (Horizon 2020), the two leading European nations in terms of this research, the UK and Germany are not closely engaged in research collaborations on the topics of internet of things, cyber-physical systems and Industry 4.0. Other findings from this study include an up-to-date overview of existing and emerging advancements in the field of I4.0, from academic literature and governmental initiatives. This combines existing literature to derive common basic terminology and approaches and to incorporate existing standards into a new conceptual mechanism for embedding AI within IoT of I4.0. In this study, cyber-physical systems are defined as connecting systems for human–computer interactions, and our research focus was on artificial intelligence that meets public acceptability, security standards, and legal scrutiny. Therefore, instead of simply presenting concepts, we presented summary tables with clear narratives showing the national frameworks that are compliant and meet public acceptability, security standards, and legal scrutiny. The summary tables categorise concepts from established frameworks, while the literature review and the bibliometric analysis categorise established scientific research data records.

The summary tables developed throughout the paper, can serve as a best practice and inform initial steps taken for design and prototype of digital twins CPSs in Industry 4.0. Such prototype would improve our capacity for a comprehensive and systematic understanding of the opportunities and threats that arise when CPS are migrated to the periphery of the internet and into local IoT networks.

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Declarations

Conflict of interest On behalf of all authors, the corresponding author states that there is no conflict nor competing interest.

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