#### CAPTURING THE REALITY OF INDUSTRY 4.0 READINESS DIMENSIONS AND INDICATORS IN A DEVELOPING COUNTRY: AN ANALYSIS OF APPLYING 14.0 IN INDONESIA

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#### ARTICLE INFO

#### ABSTRACT

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Indonesia has the lowest Industry 4.0 (I4.0) readiness in South East Asia, and uses the INDI 4.0 instrument (Indonesia Industry 4.0 readiness index), which is not as comprehensive and accurate as it could be. An Initial survey confirmed that only 56.86% of respondents agreed that INDI 4.0 accurately measured readiness in manufacturing operations. Unlike the primary 14.0 indices, INDI 4.0 lacks comprehensive 14.0 dimensions and characteristics. This paper aims to identify the dimensions and indicators of I4.0 readiness to enhance INDI 4.0 through an exploratory mixed-method research approach with a multiphase research design. The first phase consisted of a qualitative approach through a documentary review, interviews, and observations to explore the dimensions and indicators of 14.0 readiness. This phase successfully identified four dimensions that experts, academics, and practitioners validated: technology, data life cycle, I4.0 design principles, and smart factory. These dimensions were broken down into 27 indicators of I4.0 readiness, then validated again through a statistical approach before being affirmed by a survey with a Pearson's correlation > 0.361 and Cronbach's alpha of 0.987, which indicated validity and reliability. The second phase, using a quantitative approach through a survey, confirmed the four dimensions and 27 indicators for measuring 14.0 readiness levels; these were 'not ready', 'early stage', 'moderate readiness'. 'readiness ripe', and 'already implemented'. This finding comprehensively measured I4.0 readiness based on feedback from industries struggling to adopt 14.0, especially in developing countries such as Indonesia. This finding also differed from existing indices (Acatech, RAMI 4.0, Dreamy, SIMMI 4.0. IMPULS) from developed countries and industries already stable in implementing I4.0.

#### OPSOMMING

Indonesië het die laagste industrie 4.0 (I4.0) gereedheid in Suidoos-Asië, en gebruik die Indonesië Industrie 4.0 indeks (INDI 4.0) instrument, wat nie so omvattend en akkuraat is as wat dit kan wees nie. 'n Aanvanklike opname het bevestig dat slegs 56,86% van die respondente saamgestem het dat INDI 4.0 die gereedheid in vervaardigingsbedrywighede akkuraat gemeet het. Anders as die primêre I4.0-indekse, het INDI 4.0 nie omvattende I4.0-dimensies en -eienskappe nie. Hierdie artikel het ten doel om die dimensies en aanwysers van I4.0-gereedheid te identifiseer om INDI 4.0 te verbeter deur 'n verkennende gemengde-metode navorsingsbenadering met 'n multifase navorsingsontwerp. Die eerste fase het bestaan uit 'n kwalitatiewe benadering deur 'n literatuur oorsig, onderhoude en waarnemings om die dimensies en aanwysers van I4.0gereedheid te verken. Hierdie fase het vier dimensies suksesvol geïdentifiseer wat deur kundiges, akademici en praktisyns bevestig is: tegnologie, datalewensiklus, I4.0-ontwerpbeginsels en slim fabriek. Hierdie dimensies is opgebreek in 27 aanwysers van 14.0 gereedheid, dan weer bekragtig deur 'n statistiese benadering voordat dit bevestig is deur 'n opname met 'n Pearson se korrelasie > 0.361 en Cronbach se alfa van 0.987, wat geldigheid en betroubaarheid aangedui het. Die tweede fase, met behulp van 'n kwantitatiewe benadering deur 'n opname, het die vier dimensies en 27 aanwysers vir die meting van 14.0-gereedheidsvlakke bevestig; dit was 'nie gereed', 'vroeë stadium', 'matige gereedheid', 'gereedheid ryp', en 'reeds geïmplementeer'. Hierdie bevinding het 14.0-gereedheid omvattend gemeet op grond van terugvoer van nywerhede wat sukkel om 14.0 aan te neem, veral in ontwikkelende lande soos Indonesië. Hierdie bevinding het ook verskil van bestaande indekse (Acatech, RAMI 4.0, Dreamy, SIMMI 4.0. IMPULS) van ontwikkelde lande en nywerhede wat reeds stabiel is in die implementering van 14.0.

### 1. INTRODUCTION

#### 1.1. Industry 4.0

Dr Johanes Helbig, in the document Germany vision in 14.0 [1], said:

Implementing the [Industry 4.0] vision will enable employees to control, regulate, and configure smart manufacturing resource networks and manufacturing steps based on situations and context-sensitive targets. Employees will be freed up from performing routine tasks, enabling them to focus on creative, value-added activities. They will thus retain a key role, particularly in quality assurance. Simultaneously, flexible working conditions will enable excellent compatibility between their work and personal needs.

There is no consensus about a definition of the Industry 4.0 (I4.0) concept, except that it provides benefits, vision, and technology. Still, much of the literature mentions that I4.0 consists of four design principles in implementing I4.0: interconnection, information transparency, decentralised decisions, and technical assistance. It was first proposed by Mario Hermann, Tobias Pentek, and Boris Otto from Dortmund University in Germany [2]. Figure 1 shows the design principles of I4.0.

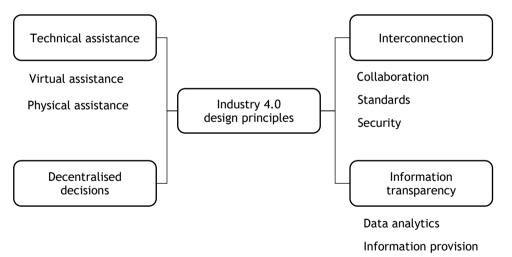


Figure 1: Design principles of Industry 4.0 [2]

'Interconnection' refers to machines, devices, and sensors connected by wireless communication through the Internet of Things (IoT). 'Information transparency' means the interconnection of objects, such that people quickly access all the information for the right people or objects, providing a virtual copy of the physical object. 'Decentralised decisions' make the interconnection of things and people more transparent in using multi-source information. 'Technical assistance' refers to virtual assistance to provide decision support systems to empower people to drive decision-making by transforming data into visualising information. Furthermore, technical assistance provides physical assistance, such as advanced robotic technology, to support a factory floor's operations.

Implicitly, these four design principles of I4.0 (Figure 1) are similar to some technological elements as the dimensions of I4.0 [3] [4] [5]:

- a) Real-time capability and system integration
- b) Big data analytics, machine learning, and artificial intelligence
- c) Decision support systems and automated decision making
- d) Vertical and horizontal system integration and cyber-physical systems

14.0 delivers the most relevant qualifying technologies frequently cited in the literature, such as big data, the Internet of Things, cloud computing, autonomous robots, additive manufacturing, cyber-physical systems, and augmented reality [6]. 14.0 has four dimensions that are affirmed in some literature reviews:

- a) Data life cycle [7] [8] [9] [10] [11] [12] [13] [14] [15] [16] It is a data journey that requires advanced technology from the source, collecting, transmitting, storing, processing, analysing, and transforming data into helpful information to support a decision.
- b) Technology elements [2] [5] [17] [18] [19] [20] [21] [22] The top seven technologies most frequently cited in the literature related to I4.0 topics [18] are the Internet of Things (IoT), big data, cloud computing, cyber-physical systems (CPS), additive manufacturing, autonomous robots, and augmented reality.
- c) Smart factory [7] [16] [19] [23] [24] [25] [26] The smart factory uses smart design, smart planning and process optimisation, smart material distribution and tracking, smart production, smart monitoring, smart maintenance, and smart quality [7].
- d) I4.0 design principles [2] [27] [28] [29] This concept is the primary reference in I4.0 research and digital transformation and is referenced in many leading papers. It consists of four I4.0 design principles: interconnection, information transparency, decentralised decisions, and technical assistance [2].

In addition, various references describe the characteristics of I4.0 in these terms:

- a) Decentralisation decisions involve the role of technology in the form of computer systems [8] [28] [29] [30] [31] [32] [33].
- b) Robot support and virtual support are available, supported by IoT technology [2] [23] [28] [29] [34] [35].
- c) Having transparent information by transforming raw data and information from the physical and virtual worlds [30] [36] [37] [38].
- d) Using I4.0 technology, such as the IoT, cyber-physical systems, cloud, big data, and artificial intelligence [2] [5] [17] [18] [19] [20] [21] [22].
- e) Various resources and work operations are interconnected quickly (even in real-time) and efficiently, and there is communication among all participants [2] [28] [29].
- f) Decision-making through transforming data into information by data analytics [7] [13] [25] [39].
- g) Vertical and horizontal integration functions are distributed throughout the network and interact across hierarchy levels [19] [27] [40] [41] [42].
- Machines, equipment, and facilities run autonomously and are self-controlling [7] [18] [27] [43] [44] [45].
- i) Employees are freed up from performing routine tasks, enabling them to focus on creative, valueadded activities [1] [6] [46] [47] [48].
- j) Dynamic business and engineering processes enable last-minute changes to production and deliver the ability to respond flexibly to disruptions and failures (agility - flexible system and machines)
   [5] [19] [49] [38] [50].
- k) Real-time capability [5] [6] [7] [14] [43].

### 1.2. Industry 4.0 readiness index and INDI 4.0

The I4.0 readiness index is an instrument to measure the readiness level for adopting I4.0. This paper aimed to identify the dimensions and indicators for measuring I4.0 readiness to enhance INDI 4.0 by converting an I4.0 concept, with its dimensions and characteristics as a variable, to be an operational variable that could be measured by exploring its dimensions and indicators. In this research, 'variable' refers to a concept, person, place, thing, or phenomenon that can be measured [51] [52]. Dimensions and indicators of I4.0 were explored from documents and then discussed and explored again through interviews; so the readiness factors were based on the realities from interviews industry, practitioners, experts, professionals, and academics involved in implementing I4.0 in Indonesia. Furthermore, the I4.0 readiness index in Indonesia was also compared with the main I4.0 readiness index, using several literature reviews, to capture the reality of the I4.0 readiness dimensions and indicators.

Despite there being no literature, reference, or I4.0 index that contains a complete and perfect instrument model that meets the needs of I4.0, this study followed the recommendations and suggestions of previous research:

- a) Increasing the practicality of implementing I4.0 in the organisation [53].
- b) Developing organisational factors or other important or specific aspects of the system in detail in I4.0 [49].
- c) I4.0 initiators are generally from developed countries, so the companies that are studied are only those that have implemented and carried out I4.0 projects [49].
- d) The research requires a different perspective for example, from industries striving for the transformation of I4.0, especially from developing countries such as Indonesia [49].
- e) Some research is too focused and oriented on technology [49].
- f) Use different tools, taking into account a more diverse number of dimensions, kinds of support, and indicators to ensure that the integrity of the developed model is maintained [18].

Based on their dimensions, some I4.0 indices focus on technology and IT [42] [49] - for example, SIMMI 4.0 and RAMI 4.0, - while others consider organisational, management, and even cultural factors such as Acatech, the Singapore I4.0 smart index, and Dreamy, even when a portion of their version is familiar with innovation and new technology as a developed country. INDI 4.0 is unlike other major indices: all of its questions are designed for a single organisation's overall general scope. Some major indices, such as the Singapore smart industry readiness index, have questions designed for an organisation's functions: its facilities, shopfloor, and enterprise levels, while RAMI 4.0, DREAMY, Acatech, and so on have all of their questions designed for other functions of an organisation, such as production, quality, research and development, logistics, services, and marketing.

INDI 4.0 only measures with instruments with narrow indicators and does not collect enough comprehensive information about readiness or the characteristics of I4.0 - at least, not as described in the main literature reviews and instruments used by other countries. For example, INDI 4.0 measures the I4.0 readiness of aspects of factory operations with only four closed and multiple-choice questions that lack comprehensiveness and accuracy [54]: 1) Where is data stored? 2) What technologies are adopted in the logistics system? 3) What is the percentage of automation in the process? and 4) What advanced system is adopted in maintenance? It was confirmed by an initial survey of the industries involved in adopting I4.0 in this study. The survey result indicated an assessment of INDI 4.0 as capturing I4.0 readiness relating to the factory operations. Only 56.86% of the respondents agreed that the current INDI 4.0 assessment was accurate. Also, in the second survey, only 42.11% of the respondents answered that INDI 4.0 reported I4.0 readiness. These results indicated a low percentage and suggested that this concern required improvement, especially in the dimensions and indicators of I4.0 readiness.

## 2. METHODOLOGY

Pragmatism is the philosophical framework underpinning the mixed methods approach to provide a set of assumptions about knowledge and inquiry. It is distinguished from quantitative and qualitative approaches, which are respectively based on the philosophies of positivism and interpretivism [55] [56]. This study used a mixed methods approach to address its research objectives, so it is associated with the pragmatic paradigm and seeks to answer the research questions correctly using a multiphase research design.

Table 1 shows that Phase 1 of the multiphase method adopted a qualitative approach through a literature and document review, interviews, and observations to explore the dimensions and indicators of 14.0 readiness. The goal of this phase was to ensure that all statements about the importance level of every indicator for each dimension were valid and reliable, besides being validated by experts. Phase 2 consisted of a quantitative approach, through a survey, to confirm the relevance of the indicators of the dimensions of 14.0 readiness.

## Table 1: The multiphase research design with research objectives, questions, and types of collected data

Research objectives	Research questions (RQ)	Data collection methods	Types of data
Phase 1: To identify the dimensions and indicators of I4.0 readiness	RQ1: What are the dimensions and indicators of I4.0?	<ul><li>a. Literature review</li><li>b. Document review</li><li>c. Focus group discussion</li><li>d. Interviews</li></ul>	Qualitative 11 respondents
Phase 2: To confirm dimensions and indicators based on reality and feedback from industries.	RQ2: What is the level of importance in respect of conformity of every indicator to dimensions to represent I4.0 characteristics?	a. Survey	Quantitative 51 respondents

Phase 1 is a Qualitative method for collecting data through interviews, focus group discussion, literature review, and document review (as noted in Table 1). Eleven interviewees had the necessary qualifications and a deep understanding, knowledge, experience, position, and involvement in I4.0 adoption projects in Indonesia. Phase 2 is a Quantitative method for collecting data through a survey. The 51 survey respondents represented industries, academics, and practitioners of I4.0. The survey was based on a non-probability and purposive sample for some reasons. The Government of Indonesia chose just twenty-five sample companies in its assessment of I4.0 readiness in April 2018 because, initially, the I4.0 project invited a few companies to participate [54]. From 2018 to 2022, only 903 companies in Indonesia assessed I4.0 readiness, with 8.52% of them at the level of maturity readiness and implementing I4.0. A self-assessment of I4.0 readiness was also part of the I4.0 project [57]. A limited number of companies in Indonesia are experienced, involved, and have deep knowledge of the I4.0 concept and its adoption.

## 3. RESULTS AND DISCUSSION

#### RQ1: What are the dimensions and indicators of Industry 4.0?

The literature review was an early step in collecting related research and selecting the most relevant of it. Next, it was necessary to identify the research gap; understand the I4.0 dimensions, principles, indicators, and characteristics; explore related research; review backgrounds, concepts, and theories; and consider the most suitable research method. In achieving the research objective - to identify the dimensions and indicators of I4.0 readiness - the study explored existing models available in the literature, such as journal papers, books, internet articles, magazines, and government documents. Reviewing and analysing such kinds of literature was done in the initial steps of the research and in each of the steps. In the first research phase (the qualitative method), one of the crucial outcomes of the literature review was the development of the framework of semi-structured questions for interviews about the scope of I4.0's dimensions or pillars that were confirmed previously by experts and practitioners with qualifications in the digital transformation field and Industry 4.0.

Table 2 shows that Phase 1 of the research consisted of semi-structured interviews to collect data to identify the dimensions and indicators of readiness for implementing I4.0 in the manufacturing industry and to classify the readiness levels. The study analysed the results of interviews about the essential indicators (Column 4 in Table 2) for adopting I4.0. The interview results were then recorded and analysed using NVivo 1.2 software.

14.0's significant factors	Responses & suggestions from respondents in interviews	Summary of indicators	References
	(RSP 01 - RSP 011)		
Data life cycle: Data sources, collection, storing, transferring, analysis, transmission, and visualisation	(RSP 01 - RSP 011) Data life-cycle is an important I4.0 factor (RSP 01, 02, 05, 06, 08, 09, 111) The most substantial factor in the data life cycle is that data flowing and transforming into information must be in real- time (RSP07) The data life cycle is not only about data collection but also about how data is processed into the final goal or information (RSP02) Data visualisation and dashboards are tools for making decisions, but the most important thing is that the data has a predetermined architectural design (RSP02) In some areas, the industry applied I4.0 in the data life cycle. For example, data obtained from each sensor is streamed to the server and designed to be analysed as a decision-making aid (RSP02, 06, 07, 09) The data life cycle is the big data platform, and it is applied in Honda Motors, for example, in predictive maintenance (RSP 09) Data is to get information quickly or to respond promptly to problems. That's why it is essential (RSP 05) Respond to processes quickly using sensors and devices for capturing data, and the data life-cycle is very important for I4.0 (RSP 05) The data life-cycle in production, for example, in production engines or machines, is implemented for preventive detection, so it is essential in I4.0 (RSP 10) To reach data-driven manufacturing, applying sensors and all that stuff also requires investing in at least	Data sources in 14.0 (sensors, IoT, RFID, social media, other devices) Data collecting in 14.0 (automatic, real-time) Data storing in 14.0 (cloud, internet network) Data transfer in 14.0 (internet, IoT, cloud, real-time technology) Data processing in 14.0 (machine learning, advanced data analytics, big data, and others) Data visualisation in 14.0 (charts, diagrams, graphs, and virtual reality in real-time)	

I4.0's significant factors	Responses & suggestions from respondents in interviews (RSP 01 - RSP 011)	Summary of indicators	References
	adequate hardware and software (RSP11)		
Design principles of 14.0: Interconnection, information transparency, decentralised decisions, technical assistance (virtual & physical assistance) Interoperability, virtualisation, real-time capability, service orientation, and modularity (14.0 can provide more flexibility, reduce lead times, be customised with small batch sizes, and reduce costs) Data integration, flexible adaptation, intelligent self-organising, interoperability, manufacturing process, optimisation, secure communication, and service orientation Integrated, adapted, optimised, service- oriented, and interoperable manufacturing processes correlated with algorithms, big data, and high technologies.	Integration (RSP01) Connectivity (RSP01, 03) Real-time (RSP01) Advanced automation (RSP01) Vertical and horizontal integration (RSP01) Physical and virtual assistance is the backbone of I4.0 (RSP03) Agree that the four things above (interconnection, information transparency, decentralisation, virtual & physical assistance) are the main components of I4.0, but under certain conditions - in particular, the decentralisation of certain strategic decisions - they require human competence (RSP 05) Not all four characteristics (interconnection, information transparency, decentralisation, virtual & physical assistance) must be applied when adopting I4.0. It depends on the context of the needs and type of industry (RSP 10)	Decision decentralisation (decentralised decisions enabled by computers or CPS, computers, embedded sensors, and systems make it possible to monitor and control the physical world independently) Interconnection (machines, devices, sensors, and humans are connected through IoT, human-machine collaboration, and machine-machine collaboration) Information transparency (data patterns from the physical world are translated by computer systems into transparent information in virtual form) Technical assistance in I4.0 (physical assistance for humans, such as robotic or physical assistance, computer system assistance to support virtual human tasks or virtual assistance)	[2] [27] [28] [29]
Smart factory: Smart planning, design, production, tracking, distribution, monitoring, maintenance, and quality Data-driven manufacturing Connected, self-optimised, transparent, agile manufacturing The ability to respond flexibly to disruptions and failures, end-to-end transparency, allowing individual customer requirements to be met, which means that even one-off items, dynamic	A smart factory is a prominent part of I4.0 (RSP 09) The smart factory consists of advanced technology such as autonomous robotics, augmented reality, monitoring and tracking, real-time automation, hazard and maintenance sensing (RSP03) The smart factory should be categorized as an I4.0 characteristic, but it also requires clarification about perceptions between I4.0 and IoT (RSP14)	Smart design in 14.0 (product design shifts towards data-driven design, data is collected from internet and IoT sources, data is abundant from product users) Smart planning (automatic production run and intelligent production optimisation) Smart maintenance (predictive maintenance, real-time technology)	[7] [16] [19] [23] [24] [25] [26]

processes enable last- minute changes to for quite a while (RSP07)	Smart monitoring (early detection of	
processes enable last- minute changes to nuclear to for quite a while (RSP07)		
Transforming data collected from sources into manufacturing intelligence for decision makingfactory is first about the accuracy of the data to reduce or eliminate error or data bias to get accurate information (RSP02)acc conditional accuration diada to get accurate information (RSP02)Smart working, smart product, and smart supply chain (base technology: big data, IoT, cloud, 	abnormality, high flexibility, precision, accurate, automatic correction) Tracking system and distribution in 14.0 (real-time, up-to-date inventory, global positioning system (GPS), automatic guided vehicle (AGV), technology tracking system) Smart quality (various data-driven quality control techniques, data sources from sensors, RFID, and other machine applications)	

14.0's significant factors	Responses & suggestions from respondents in interviews	Summary of indicators	References
	(RSP 01 - RSP 011)		
	company's master plan (RSP 10)		
	Some automation approaches (karakuri sometimes go back to classical technology - for example, technology based on gravity. Classical mechanical technology is good as long as it is practical and results in the same outcome as advanced technology (RSP 10). Sometimes classical technology is good, as long as it is applicable and has a significant impact, like I4.0		
Technology: IoT	5G connectivity is a part of I4.0 technology (RPS03)	Internet of Things technology	[2] [5] [17] [18] [19]
Big data analytics Cyber-physical systems, IoT, cloud, real-time capability System Integration Machine learning Artificial intelligence Decision support system Automated decision making Vertical and horizontal integration, additive manufacturing, augmented reality, automated robots Smart working, smart manufacturing, smart product, and smart supply chain (base technology: big data, IoT, cloud, analytics)	Technology is only a key enabler but not the main factor in 14.0 (RPS03) Artificial intelligence, automation, and robotics will make the shift to a disruptive era a significant factor of 14.0 (RPS03, RSP14) Big data and IoT are identified as technology features that emerged in the 14.0 era (RPS03, RSP02, RSP06) Robots are also crucial in 14.0 technology, followed by the rapid development of drone technology in various industrial fields (RSP06) 14.0 also gives rise to interconnection and integration between humans, machine and machine to machine, computer support, advanced technology, artificial intelligence, machine learning, and others (RPS03) Various tools, technology, machines, and equipment that are identical to 14.0 are useful and powerful things but need to be adapted to existing business requirements (RSP08) For example, in Astra Daihatsu, several processes have	Big data analytic technology Cloud computing technology Cyber-physical system (CPS) technology Machine learning technology Artificial intelligence technology Decision information system (DCS) technology Automatic decision- making technology Real-time technology Vertical integration system technology (interconnection from operational or field level to top management level or business level) Horizontal integration system technology (interconnection in supply chain hierarchy from suppliers to customers)	[20] [21] [22]
	several processes have implemented artificial intelligence, such as robotic systems in welding and		

14.0's significant factors	Responses & suggestions from respondents in interviews	Summary of indicators	References
	(RSP 01 - RSP 011)		
	automatic assessment of whether the welding process results are up to standard. It is part of artificial intelligence (RSP07, RSP03)		
	Adopting 14.0 does not mean that all machines or equipment require 14.0 technology. It must be assessed according to context and needs as long as it meets and exceeds all customer requirements such as specifications, capacity, production tack time, technological capability, investment, and position of competitive advantage still under control (RSP 05, RSP 09)		
	Technology is very important in I4.0, but people must be prepared (RSP 09)		
	Toyota may not lead I4.0 technology, but in Indonesia, Toyota plays a significant role in developing local industrial technology, has I4.0 roadmap guidelines from the mother company, and plays an active role in the Indonesian government's I4.0 project (RSP 10)		
	In Astra Honda Motor, machine automation has been implemented, which records and provides real-time information about engine conditions and current abnormalities involving I4.0 technology, such as cloud and IoT (RSP 11)		
	Approving and confirming that technologies such as big data, cloud, IoT, CPS, AI, additive manufacturing, and others are one of the hallmarks of I4.0 technology and have a solid positive impact if they can be combined on-site and then can be implemented end-to-end as needed (RSP 11)		
14.0 readiness level	I4.0 levels of readiness: already implemented, highly ready, ready, less prepared, not ready (RSP 06, 08, 09, 10, 11)	Level 0 = Not ready Level 1 = Early readiness stage	[53] [58] [59] [60]

I4.0's significant factors	Responses & suggestions from respondents in interviews (RSP 01 - RSP 011)	Summary of indicators	References
	Astra Honda Motor is ready for 14.0 implementation, even though some processes still use old technology, but gradually following the mother company in the 14.0 plan (RSP 08)	Level 1 = Moderate readiness Level 3 = Readiness ripe Level 4 = Already implemented	
	The latest result of the INDI 4.0 score is 3.7 out of 4 (some sub-organizations of National Electric Company), so the company position is highly ready. (RSP06)		
	The I4.0 readiness level is at the third level (RSP 09)		
	Still ongoing I4.0 project (RSP 09, RSP 11)		
	Mitsubishi Motor Indonesia obtained a 3.2 out of 4 score (RSP 05)		
	Toyota Motor Indonesia in Sunter is already implementing I4.0 (RSP 10)		
	Toyota Motor Indonesia is one of the largest companies, but the first and second supplier levels are not ready. It must be stable first and avoid variability of product quality (RSP 10)		
	The company of respondent work for has already implemented I4.0 (RSP 11)		

Table 2 shows a summary of the indicators (column 4) based on a summary of the interviews about the characteristics of 14.0 and its indicators. This phase identified four dimensions and 27 indicators. All of these indicators were summaries derived from interviews, which were analysed manually by looking at the verbal answers of the respondents or interviewees and analysing them with QSR NVivo12 software to organise, integrate, code, and analyse the data. The main outcomes from the qualitative interviews were four dimensions and 27 indicators:

- a) Data-life cycle category (defined by six indicators, No. 1 to 6)
- b) Design principle of the I4.0 category (defined by four indicators, No 7 to 10)
- c) Smart factory (defined by six indicators, No.11 to 16)
- d) Technology category (defined by eleven indicators, No. 17 to 27)

RQ2: How the level of importance in terms of conformity of every indicator to dimensions to represent 14.0 characteristics

Table 3 shows the survey's results, confirming the 27 leading indicators; the respondents gave their opinions on how important the levels of these indicators were to these four I4.0 dimensions (Columns 4, 6, 8, 10). Column 11 is the approval level for the 27 indicators. A six-point Likert scale was used that reflected the degree of importance of every leading indicator:

- a) Very high = 6
- b) High = 5

- c) Moderate = 4
- d) Low = 3
- e) Very low = 2
- f) Not suitable = 1

For analysing the results, this study used a scale for the median values that reflected the level of importance, as follows:

- a) 5.51 6.00 = Very high
- b) 4.51 5.50 = Important
- c) 3.51 4.50 = Moderate
- d) 2.51 3.50 = Low
- e) 1.41 2.50 = Very low
- f) < 1.40 = Not suitable

The median is the most informative measure of the central tendency for skewed distributions or distributions with outliers. For example, the median is often used to measure the central tendency for income distributions, which are generally highly skewed. Because the median only uses one or two values, it is unaffected by extreme outliers or non-symmetric distributions of scores. In contrast, the mean and mode can vary in skewed distributions.

# Table 3: Level of importance of conformity of 14.0 indicators to 14.0 dimensions and approval of 14.0 readiness indicators

		Importance level of indicators to I4.0 dimensions								Approval level of	
No	Leading indicators of I4.0	Data life- cycle		Smart factory		Technology elements		Design principle		indicators to 14.0 readiness	
		Avr	Med	Avr	Med	Avr	Med	Avr	Med	Median	
1	Data sources in I4.0 (sensors, IoT, RFID, social media, other devices)	5.45	6	5.31	5	5.35	5	5.37	5	5	
2	Data collecting in I4.0 (automatic, real-time)	55	6	5.33	5	5.29	5	5.39	5	5	
3	Data storing in I4.0 (cloud, internet network)	5.49	6	5.37	6	5.53	6	5.45	6	5	
4	Data transfer in 14.0 (internet, IoT, cloud, real-time technology )	5.51	6	5.39	5	5.43	5	5.51	6	5	
5	Data processing in 14.0 (machine learning, advanced data analytics, big data, and others)	5.43	5	5.43	6	5.49	6	5.39	5	5	
6	Data visualisation in 14.0 (charts, diagrams, graphs, and virtual reality in real-time)	5.22	5	5.18	5	5.18	5	5.31	5	5	

	Leading indicators of I4.0	Impor	tance	level o	f indica	ators to	14.0 di	mensio	ns	Approval level of
No		Data life- cycle		Smart factory		Technology elements		Design principle		indicators to 14.0 readiness
		Avr	Med	Avr	Med	Avr	Med	Avr	Med	Median
7	Smart design in I4.0 (product design shifts towards data-driven design, data is collected from Internet and IoT sources, data is abundant from product users)	5.29	5	5.25	5	5.25	5	5.27	5	5
8	Smart planning (automatic production run and intelligent production optimisation)	5.41	5	5.37	6	5.31	5	5.33	5	5
9	Smart maintenance (predictive maintenance, real-time technology)	5.35	5	5.39	5	5.41	6	5.33	5	5
10	Smart monitoring (early detection of abnormality, high flexibility, precision, accuracy, automatic correction)	5.53	6	5.47	6	5.39	6	5.33	5	5
11	Tracking system & distribution in 14.0 (real- time, up-to-date inventory, global positioning system (GPS), automatic guided vehicle (AGV), technology tracking system)	5.51	6	5.43	5	5.31	5	5.35	6	5
12	Smart quality (various data-driven quality control techniques, data sources from sensors, RFID, and other machine applications)	5.39	5	5.41	5	5.29	5	5.27	5	5
13	Internet of Things technology	5.55	6	5.43	6	5.53	6	5.43	6	5
14	Big data analytic technology	5.55	6	5.41	5	5.43	5	5.35	5	5
15	Cloud computing technology	5.41	5	5.39	5	5.45	5	5.33	5	5
16	Cyber-physical system (CPS) technology	5.20	5	5.22	5	5.16	5	5.24	5	5
17	Machine learning technology	5.24	5	5.31	5	5.31	5	5.16	5	5
18	Artificial intelligence technology	5.20	5	5.31	5	5.29	5	5.25	5	5

		Impor	tance	level o	f indica	ators to	14.0 di	mensio	ns	Approval level of	
No	Leading indicators of I4.0	Data life- cycle		Smart factory		Technology elements		Design principle		indicators to 14.0 readiness	
		Avr	Med	Avr	Med	Avr	Med	Avr	Med	Median	
19	Decision information system (DCS) technology	5.24	5	5.29	5	5.18	5	5.22	5	5	
20	Automatic decision- making technology	5.27	5	5.12	5	5.20	5	5.12	5	5	
21	Real-time technology	5.43	5	5.45	6	5.41	6	5.41	6	6	
22	Vertical integration system technology (interconnection from operational or field level to top management level or business level)	5.31	5	5.24	5	5.14	5	5.20	5	5	
23	Horizontal integration system technology (interconnection in supply chain hierarchy from suppliers to customers)	5.25	5	5.20	5	5.22	5	5.14	5	5	
24	Decision decentralisation (decentralised decisions enabled by computers or CPS, computers, embedded sensors, and systems make it possible to monitor and control the physical world independently)	5.16	5	5.22	5	5.27	5	5.10	5	5	
25	Interconnection (machines, devices, sensors, and humans are connected through IoT, human-machine collaboration, and machine-machine collaboration)	5.37	5	5.49	6	5.39	5	5.35	5	5	
26	Information transparency (data patterns from the physical world are translated by computer systems into transparent information in virtual form)	5.31	5	5.24	5	5.29	5	5.22	5	5	

No		Importance level of indicators to I4.0 dimensions								Approval level of
	Leading indicators of I4.0	Data life- cycle		Smart factory		Technology elements		Design principle		indicators to 14.0 readiness
		Avr	Med	Avr	Med	Avr	Med	Avr	Med	Median
27	Technical assistance in 14.0 (physical assistance to humans such as robotic or physical assistance, computer system assistance to support virtual human tasks or virtual assistance)	5.14	5	5.16	5	5.25	5	5.20	5	5

There are 27 leading indicators of 14.0 in four dimensions in Table 3. All indicators obtained a median from 5 to 6 (columns 4, 6, 8, 10) as results from the survey. It meant that most indicators were relevant because the median levels ranged from 4.51 to 5.50 = important and from 5.51 to 6.00 = very important. Also, Column 1 shows the approval level of the respondents for the 27 indicators as 'important', with a median of 5. Only indicator No. 21, real-time technology, with a median of 6, was 'very important' as an 14.0 readiness indicator. The instrument was also validated again through a statistical approach before being affirmed by a survey with a Pearson correlation > 0.361 and Cronbach's alpha = 0.987 > 0.6, which meant it was valid and reliable.

#### 4. CONCLUSION

Based on the literature review, I40 existing indices, and related documents, Industry 4.0 comprises four dimensions - data life-cycle, technology, I4.0 design principles, and smart factory - broken down into 27 indicators. It was confirmed by a survey of the industry and stakeholders involved in the I4.0 project; thus, it was aligned with reality and with the implementation practices associated with the indicators of I4.0 readiness. All of the dimensions and indicators. These findings could enhance INDI 4.0, which assesses an entire organisation in general terms, with general questions and some capturing of the comprehensive dimensions, characteristics, and indicators of I4.0 readiness, as described in the literature review.

The interviews revealed other dimensions, such as management, people, and smart maintenance. Smart maintenance has been included in the smart factory. At the same time, the dimensions of management and people, in general, are generally required by other concepts, apart from I4.0, because they are not unique to I4.0. In addition, this study only focused on I4.0 in the factory operation area. Further research would need to explore other dimensions to complement and improve the accuracy in measuring I4.0 readiness, especially those that are important for I4.0 readiness.

The major I4.0 indices (Acatech, IMPULS, RAMI, DRAMY, SIMMI 4.0), which are widely adopted around the world, are all from developed countries [18] [61] [62] [63], which have many different variables from those of developing countries such as Indonesia, including culture, people, priorities, industry characteristics, starting points, challenges, opportunities, politics, and so on. This study offers a more comprehensive instrument, especially for industries that are struggling to adopt I4.0. This study has done what previous research recommended: exploring more perspectives to complete the existing indices [18] [49]. Further research is expected to do the same.

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