



RFID in Industry 4.0: The technology's role in Shop Floor Control's Operationalisation

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Abstract. The advent of Information and Communication Technology brought some advances in the field of Operations Management. The RFID, when used as Ordering Coordination System provides benefits not only for Production Control. To identify these benefits, this work used a Systematic Literature Review. The results presented 15 benefits, classified into 4 areas that impact the Shop Floor Control and a framework containing the information flow necessary to operationalise the SFC 4.0. The discussion presented some insights relating the benefits with the maturity level of the entire SFC ecosystem when adopting RFID technology and a brief agenda for future studies in this field of knowledge. It was possible to conclude that although the use of RFID is not new, its use as a premise for SFC operationalisation is yet in its initial stage of maturity, which could be reinforced by the fact that the most cited benefits were related to the initial stage of informational flow presented in the framework.

Keywords: Industry 4.0; RFID; Manufacturing Planning and Control; Shop Floor Control; Systematic Literature Review.

1 Introduction

The advent of technology, changes in marketing requirements, and the evolution of Information and Communication Technologies (ICT) have brought some challenges to organisations, grounding the natural development of the Strategic Paradigm of Manufacturing Management (SPMM). This evolution, however, got complexity for Shop Floor Control due to the Work In Process and Batch Size reduction, highly customised products required in a short period, with high-quality standards. [1] pointed out that Shop Floor Control Systems (SFCS) was developed to overcome these problems. According to [2], the main issues that this approach seeks to solve are related to productivity, improved management of SF information, reduced complexity in terms of information handling, and facilitated decision-making process. For [1], Shop Floor Managers face unpredictable risks in day-to-day operation, such as errors, defects in the supplies of components, failures, and machines breakdown. The problem of these occurrences in 4.0 environments depends on the dynamicity that evolves these environments. In traditional paradigms, the risks or events are reported only days or even

monthly [3]. This fact imposes high inertia to solve operational problems related to SFC. Notwithstanding these issues, organisations face fluctuation in the demand and lead times, making a planned production control strategy unreliable. The use of the Radio Frequency Identification (RFID) technology as Ordering Coordination System (OCS) for Industry 4.0 was proposed by [4] to operationalise Shop Floor Control 4.0. Despite this proposal and the high number of studies considering RFID application, not only in manufacturing environment but also in the entire Supply Chain Management, to the best of our knowledge, no work systematically collects and analyses the benefits of RFID adoption as OCS for SFC in 4.0 environments. To fill this gap is just the objective of this research. Therefore, we use a SLR to collect and critically analyse the existing knowledge in this field of knowledge. To address the proposed objective, this study has three interrelated questions:

- 1– *What are the benefits of RFID technology in SFC 4.0?*
- 2– *How do these benefits impact (allow the operationalisation) of the SFC 4.0?*
- 3– *What does the identified state of the art say about new research opportunities?*

Thus, the remainder of this paper is organised as described: Section 2 presents the theoretical background. Section 3 contains the methodological aspects. Section 4 shows the results. Section 5 discusses the results proposing a brief research agenda, and Section 6 offers the conclusion, inferring about theoretical and practical implications of this research.

2 Theoretical Background: Shop Floor Control and RFID

2.1 Industry 4.0, Shop Floor Control and RFID, an Integrated Approach

The complexity and dynamics of the manufacturing environment are growing due to the changes in manufacturing demand, requiring some types of products, small lot sizes, and short lead-time to market [5]. To meet these new requirements, the Industry 4.0 SPMM, which, in its essence, is composed of disruptive technologies responsible for giving “intelligence” to the cyber part of a physical entity, has data acquisition as a premise. Strongly interrelated with the Fourth Industrial Revolution, RFID is an important, source of Shop Floor Control (SFC) operationalisation in a 4.0 environment. To summarise, Industry 4.0 utilises technology such as Cloud Computing, Big Data and Analytics and Digital Twin to evolve the machinery from automation to autonomy. Considering this evolution, the autonomy level is not acquired by a single machine but by the entire system, called Machine to Machine communication (M2M Communication).

Concerning SFC, we adopt the Gartner definition, for whom SFC is a system of computers and controllers tools used to schedule, dispatch and track the progress of work orders through manufacturing based on defined routings [6]. This area is one of the most challenging for manufacturing managers, given the infinity possibility of disruptions that involve machinery, operators, raw material and others. For each set of volume-variety combinations, there is an OCS that better fits the Manufacturing system’s objective. In Industry 4.0, a consensus is being established about RFID as an OCS [7].

Regarding RFID, [8] points to this technology as the primary source of data generation in intelligent environments, such as Smart Factories. Being found in standard and long-range formats, the RFID, according to [8], has the fastest recognition speed and can read tags up to 15 meters, being suitable for factory usage. When IoT and RFID are applied to the manufacturing process, heterogeneous RFID-based data, real-time and substantial data, are generated and recorded, called industrial Big Data [9]. Big Data is an Input for Analytics to treat and identify unknown relationships among data, which provide feedback to the system, adjust manufacturing and machines parameters, and serve as a basis for decision making.

3 Research’s Methodological Sequence

This study is theoretical-conceptual and presents a literature review on RFID applied to SFC 4.0 as the first attempt to show and explain the benefits of this technology for operations, proposing a literature classification and analysis. [10] points that a SLR is an essential endeavour by itself and not merely a review of previous writing. Moreover, it responds to specific questions and is a “methodology that locates existing studies, selects and evaluates contributions, analyses and synthesises data, and reports the evidence to allow reasonable, clear conclusions about what is and is not known [11].

In addition, this SLR takes the three steps proposed by [12], comprised of Planning, Conduction and Dissemination. The Planning step is summarised in the research protocol (Table 01). The defined parameters were applied in the conducting stage, and the scanning process was then started. In the dissemination stage, the data were treated, and results were presented. The step-by-step is detailed in the following subsections.

3.1 Planning Stage

The research criteria are summarised in the following table.

Table 01: Research Protocol

Research Protocol	
Objective	To identify the benefits of RFID technology for Shop Floor Control in Industry 4.0 environments;
Guiding Questions	1 – <i>What are the benefits of RFID technology in SFC 4.0?</i> 2 – <i>How do these benefits impact (allow the operationalisation) of the SFC 4.0?</i> 3 – <i>What does the identified state of the art say about new research opportunities?</i>
Database	<i>Engineering Village, Scopus, Web of Science</i>
Period	<i>From 2011 to 2022</i>
Document type	<i>Journal Articles, Articles In Press, Conference Papers, Review</i>
Language	<i>English</i>
((Data Ingestion AND Production Control) OR (Data Flow AND Production Control) OR (Data Collection AND Production Control) OR (Database AND Production Control) OR (Data Acquisition AND Production Control) OR (Data Processing AND Production Control) OR (Data Processing AND Production Control) OR (Data Management AND Production Control) OR (Smart Support AND Production Control) OR (Data Driving Simulation AND	

Production Control) OR (Data Driving Simulation AND Production Control) OR (Data Driving Simulation AND Production Control) OR (Data Driving Production Control) OR (Data-Driven AND Production Control)) AND ((shop floor) OR (Order Progress) OR (Order Monitoring) OR (Order Releas*) OR (Production Scheduling) OR (Production Sequencing) OR (Production Control) OR (Production Scheduling)) AND ((Industry 4.0) OR (Industrie 4.0) OR (Fourth Industrial Revolution) OR (Smart Factor*) OR (Smart Production) OR (digitali?ation) OR (digitali?ation) OR (Virtual factor*) OR (digital factor*) OR (data driven production) OR (real time production) OR (data collection) OR (machine learning) OR (Shop Floor) OR (data acquisition) OR (Data processing) OR (order progress) OR (Order release) OR (order monitoring))	
Criteria	Criteria Explanation
Exclusion (NR)	NR-1: Articles considering other disruptive technologies; NR-2: The definition of “RFID” is not related to the Operational Environment;
(LR)	LR: Articles that cite RFID but do not express discussion about the role of this technology in the 4.0’s Shop Floor Control
Inclusion (PR)	PR: Articles that focus on one or a few benefits/barriers of RFID
(CR)	CR: Articles featuring RFID studies, pointing out benefits and their relationship with internal organisational functions;

Source: The authors

To screen the papers, we utilise the Software StArt. The screening process adopted is summarised in figure 1.

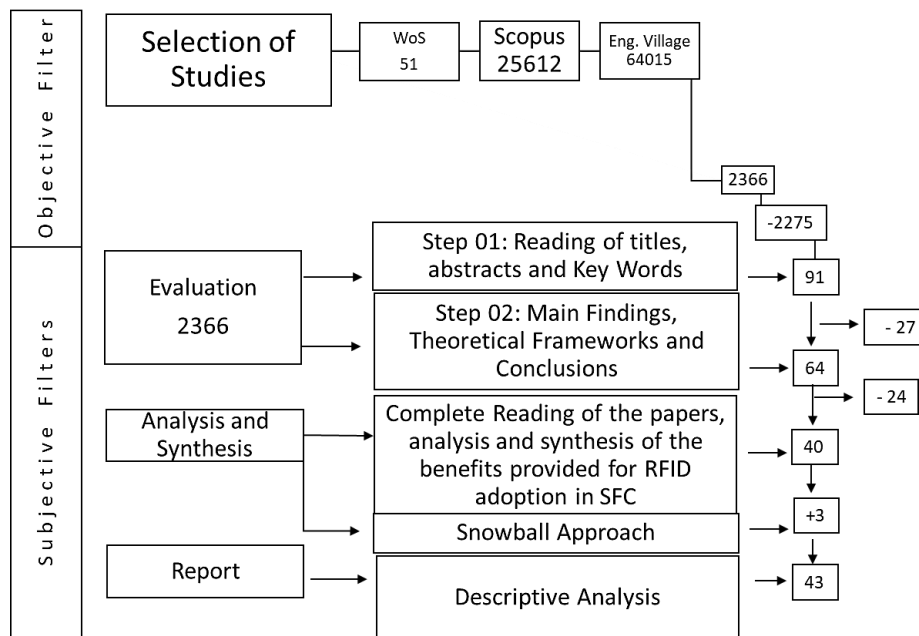


Figure 1: Summary of article’s selection

4 Research Results: The Dissemination Stage

This research identified 43 papers that pointed out the benefits of RFID adoption for SFC 4.0 operationalisation. Figure 02 shows the number of documents published by year. Analysing this figure, we can see that the number of research increases in the last years, which means that this field of study is growing.

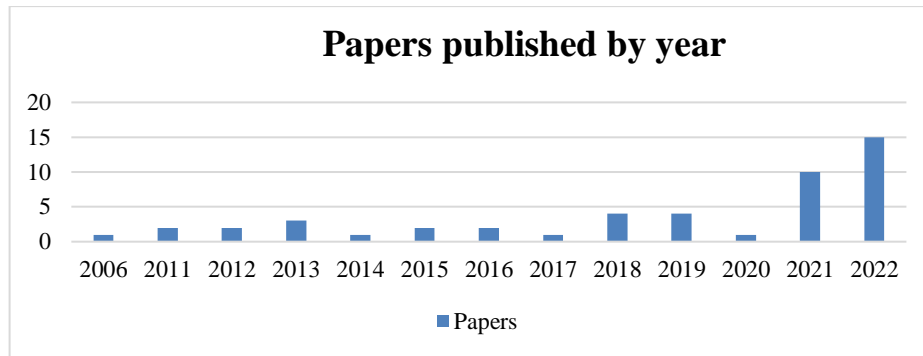


Figure 2. Papers published by year

Regarding the results, table 02 presents the benefits according to their respective area of impact. Maintenance Management was responsible for 06 benefits (24%), Quality Management was responsible for 04 benefits (16%), the same percentage as the Financial Impacts group. The last group, Operations Management, were responsible for 11 benefits (44%), being the most impacted area. The table also shows a code and description of each capability. This code allows us to insert the capabilities in the software NVivo 11, verifying the existence of any semantical similarity. The last row, containing the authors, shows many times the ability was pointed, considering the sample analysed. Of the 27 Benefits, the most cited one was Real-Time Monitoring (pointed out by 62,80% of the authors. The second most cited benefit is Decision-Making optimisation. This benefit was cited by 18,60% of the authors. The third benefit with more studies is the Flexible Real-time Shop Floor Scheduling, with 16,29% of the studies. To better illustrate the objective of this research, a framework containing the informational flow between the physical and the cyber elements of an ecosystem for SFC 4.0 is above presented. Figure 3 describes the informational flow between physical and cyber elements that compose an ecosystem 4.0 for Shop Floor Control. In this framework, the RFID, responsible for data-gathering, collects and uploads information for Big Data. An Analytics threats the data and identifies the relationship between variables. These relationships and opportunities for continuous improvement are tested in a Digital Twin, and if any gain is identified, the parameter is replaced with the physical machine.

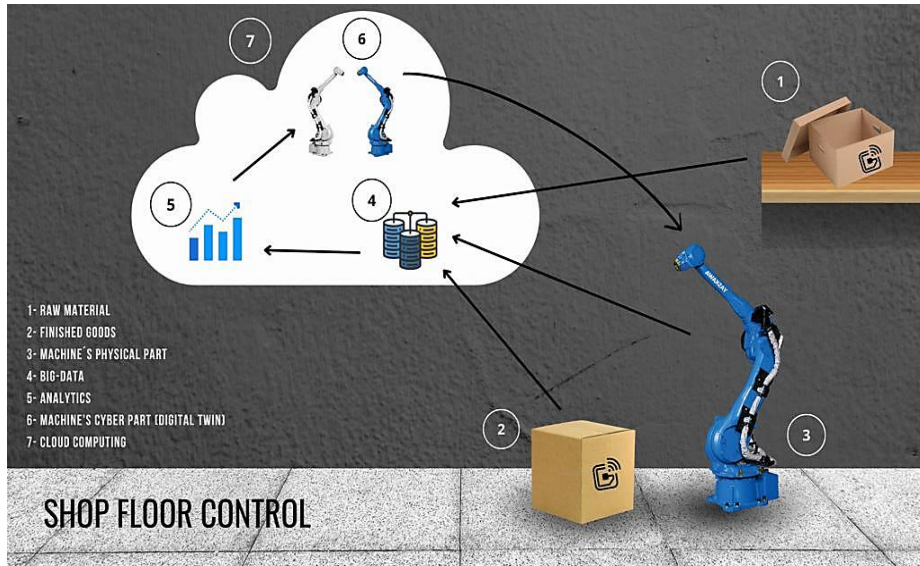


Figure 3. Informational Flow between Physical and Cyber resources in SFC 4.0

Table 2. Benefits provided for RFID adoption for SFC 4.0 operationalisation

Benefits	Code	Description	Authors
Maintenance Management			
Allow Predictive, Reactive, Preventive and Prescriptive Maintenance	MM1	Real-time data gathering allows the adoption of diverse maintenance approaches	1, 17, 22, 24, 41, 42
Control Systems' Reliability and Security	MM2	Data gathering helps to improve the Security and Reliability of the Control Systems	8, 17, 20
Reduced Unplanned Downtime	MM3	Down-time identification capability is acquired as a result of the Improved Fault Prediction	17, 20, 22, 25
Development of Digital Maintenance	MM4	Maintenance can be remotely executed in the product through Digital Twin and replicated to the physical machine	17
Allow assessment of maintenance impacts on manufacturing	MM5	RFID allow managers to assess the impacts of the maintenance in the manufacturing system	17
Fault Prediction Capability is Improved	MM6	Fault Prediction Capability improves the machines' Downtime, reflecting in better asset use	2, 3, 16, 20, 22
Quality Management			
Zero Defects Manufacturing	QM1	To produce beings without any defect	1, 8, 28
Real-Time Monitoring / Ongoing Quality Management	QM2	Real-time monitoring allows ongoing Quality Management and Problem Identification	2, 3, 7, 12, 15, 16, 17, 19, 20, 21, 23, 24, 25, 26, 28, 31, 32, 33, 34, 35, 36, 37, 38, 39, 40, 43
Increased Production Efficiency through Quality Prediction	QM3	Monitoring the Machines' status related to the most diverse variables allow Quality Problems Prediction	2, 10, 15, 16, 22, 31
Improve Operational Efficiency and Productivity	QM4	The most diverse source of data, real-time updated, allows improved production scheduling and mac efficiency	5, 21, 25, 32
Financial Impacts			
Reduced Production Costs	FI1	Through improved accuracy of data, the production cost is reduced	1, 2, 5, 24
Minimise machine repair costs for unexpected failures	FI2	Through SFC rescheduling, repair costs may be minimised	17, 24
Investment in capital is reduced	FI3	RFID collects data that can allow the organisation to identify the better assets to invest	17
Maximise Value of Limited Resource	FI4	The most diverse benefits of RFID adoption, such as the one related to Quality, Process and Maintenance issues, allow maximisation of machines' financial value	14, 24



Operations Management			
Allow End-to-End 4.0 Functionalities	OP1	The entire benefits of I-4.0 are strongly dependent on the data-gathering quality	1, 25
Allow Digital Twin Implementation	OP2	The digital twin technology's fidelity depends on the amount of collected data	1, 7, 25, 27
Real-Time Process Inference	OP3	Managers can react fast in case of production problems once the manufacturing status is being on time managed	3, 19, 21, 23
Decision-Making Optimization	OP4	Making use of Big Data, the adoption of Analytics can identify relationships among variables not clearly identified by the human rationality	4, 5, 11, 24, 25, 28, 36, 38
Reduced Communication Overhead and Algorithm Delay	OP5	The Algorithm Overhead is Reduced through real-time data gathering, which can constantly update its status.	14, 23, 25, 27
Machine Learning Approach is assessed through Data Gathering	OP6	RFID is a source of real-time data gathering that, when used by analytics and replied to a Digital Twin, allows Machine Learning	19, 25
Integrated Production and Maintenance Scheduling	OP7	Data collected from product, production and machines parameters allow integrated production and maintenance scheduling	17, 24
Flexible Real-Time Flow Shop Scheduling	OP8	Through real-time monitoring, the status of the machine can be assessed, and the production scheduling be planned or replanned in case of any unexpected occurrence	4, 18, 19, 21, 23, 29, 30
Idle-Time Minimization in Low Volume High Variety Manufacturing Environment	OP9	The idle time can be minimised through improved SFC	21
Capacity loss of a bottleneck is improved	OP10	Real-time data gathering provided for RFID allows the use of constraints theory principles to manage bottleneck resources	21
Setup time is dramatically improved	OP11	Setup times can be minimised or even reduced to practically zero with the use of RFID	23

Sources: 1. [12]; 2. [13]; 3. [14]; 4. [15]; 5. [16]; 6. [17]; 7. [18]; 8. [19]; 9. [20]; 10. [21]; 11. [22]; 12. [23]; 13. [24]; 14. [25]; 15. [26]; 16. [27]; 17. [28]; 18. [29]; 19. [30]; 20. [31]; 21. [32]; 22. [33]; 23. [34]; 24. [35]; 25. [36]; 26. [37]; 27. [38]; 28. [39]; 29. [40]; 30. [41]; 31. [42]; 32. [43]; 33. [44]; 34. [44]; 35. [45]; 36. [46]; 37. [47]; 38. [48]; 39. [49]; 40. [50]; 41. [51]; 42. [52]; 43. [53];

5 Discussion

RFID technology is the main point of data gathering in SFC 4.0. Soon, it's a premise for SFC 4.0 operationalization. Analyzing the framework (Figure 3), is it possible to perceive that the informational flow that walks throughout the SFC, from raw material to the feedback point, in the Digital Twin, pass to 7 phases. Considering that the more phases organizations has, the higher is the maturity level of its SFC 4.0. Building a parallel with the benefits presented in table 2, we can perceive that the benefits with highest number of citation is related to the initial stages of implementation of the ecosystem 4.0, as real-time monitoring and decision making optimization. Benefits that directly depends on high maturity level as the development of digital maintenance, didn't receive a high number of citation. Thus, we propose a research agenda, according to the four areas where the benefits were previously classified:

- **Maintenance Management:** Although data gathering allow diverse maintenance approaches as presented in table 2, some researches could to focus on the integration of different approaches and its impacts in the main Key Process Indicators of the maintenance area;
- **Quality Management:** With the autonomy level acquired with a complete SFC 4.0 ecosystem, the quality managers and staff's role needs to be studied;
- **Financial Impacts:** Researches focus on internal cost minimization, however, future researches could focus on the implementation cost to build a SFC 4.0;
- **Operation Management:** Researches could to focus on the development of the final stages of SFC 4.0 as, the stages that allows feedbacks for a complete loop evolving an autonomy continuous improvement system;
- **Integrated approaches as ISM/Fuzzy MICMAC** could to hierarchize the benefits, in order to provides a roadmap for RFID adoption as OCS for SFC 4.0;

6 Conclusions

The RFID technology is not new, but its use as a premise for SFC 4.0 operationalisation is yet in the initial stage. Taking part in this work, it was possible to identify and classify the benefits provided by adopting RFID technology in SFC as a premise for Smart Shop Floor Control. These benefits were scattered in the literature and cited according to researchers' interests. After systematically reviewing the selected papers, state of the art was identified, and a plan for future research was then proposed. This research, like any other, has its limitations, some of them related to the SLR method. We hope this study could be seen as a source for many others derived from the results here presented.

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