Proceedings of the Design Society, Volume 5: ICED25
https://doi.org/10.1017/pds.2025.10228

Investigating data-driven requirements management for smart PSS development

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ABSTRACT: The development of smart Product-Service Systems (smart PSS) introduces unique challenges for requirements management due to their dynamic, data-driven, and multidisciplinary nature. This paper investigates methods and principles for data-driven requirements management, emphasizing lifecycle alignment, data utilization, multidisciplinary collaboration, and customer-centricity. A systematic literature review forms the basis for assessing 16 existing frameworks, focusing on their suitability for the data-driven design of smart PSS. Key gaps are identified in areas such as data planning, lifecycle integration, and the handling of system-level requirements. To address these challenges, this study proposes principles for data-driven requirements management that leverage real-time data, ensure traceability, foster interdisciplinary alignment, and adapt to contextual variables.

KEYWORDS: product-service systems (PSS), systems engineering (SE), design methods, specification, data-driven design

1. Introduction and problem clarification

Smart Product-Service Systems (smart PSS) integrate physical products with digital services, connected through Information and Communication Technologies (ICT). This convergence of products and services enhances customer engagement, enables personalized experiences, and offers long-term value by enabling adaptation to meet users' evolving needs. The shift toward smart PSS is driven by advancements in sensors, computing power, data communication, and digitalization, allowing companies to meet customer demands for customized and adaptive products and services (Wang et al., 2019; Wiesner et al., 2017a). Unlike traditional products or services, smart PSS require holistic approaches that consider interconnected product-service ecosystems extending beyond the initial purchase with the aim of building lasting customer relationships through continuous engagement throughout the lifecycle (Paliyenko et al., 2023; Schweitzer, 2010).

As smart PSS become more prevalent, traditional product or service development methodologies reveal inherent limitations in handling their unique requirements (Paliyenko et al., 2024a). These systems involve dynamic interactions between multiple disciplines, such as mechanical engineering, software development, and service design, all of which have distinct terminologies, goals, and approaches. The interplay of these disciplines complicates requirements management (RM), increasing the risk of misalignment and inefficiency (Paliyenko et al., 2023). Additionally, smart PSS development demands a robust RM approach that incorporates data-driven insights and feedback from later lifecycle stages (e.g., use, maintenance) back into early development phases to ensure adaptability and responsiveness to customer needs (Paliyenko et al., 2023; Aurich et al., 2019). Consequently, there is an evident need for suitable methods in smart PSS RM that capture the dynamic, interconnected nature of these systems and address the multidisciplinary demands inherent to their development. As current RM methods fall short when applied to smart PSS, they often lack the flexibility to adapt to continuous, lifecycle-driven data input and do not integrate feedback loops from commissioning and operations back into the development

https://doi.org/10.1017/pds.2025.10228 Published online by Cambridge University Press

phases (Paliyenko et al., 2024b). As smart PSS increasingly provide real-time data on user behaviors and environmental factors, traditional RM methods are insufficient for capturing and leveraging this data effectively (Aurich et al., 2019; Zhou et al., 2023).
Furthermore, the transition from human-centric to data-driven RM introduces challenges in defining and prioritizing stakeholder needs. Existing methods often consider product and service requirements in isolation, overlooking their interdependencies (Paliyenko et al., 2024a). In smart PSS, this fragmented approach results in disconnected development efforts, reducing potential synergies between the product and service components and hindering alignment with customer expectations. Consequently, methods for RM must be capable of capturing and processing evolving, interlinked requirements that span various stakeholders and disciplines to foster smart PSS development (Wiesner et al., 2017a). Therefore, this research aims to investigate and propose data-driven methodologies for RM that are suitable for smart PSS development. By examining existing approaches and identifying gaps in current methodologies, this paper seeks to propose principles for data-driven RM, ensuring that requirements are dynamically aligned with real-world usage patterns and customer feedback. This paper therefore aims to answer the following research questions: *What approaches exist for requirements management in smart PSS development, and what characteristic principles should be addressed to leverage data in smart PSS?*

2. Approach to research

This research employs a structured methodology, that includes comprehensive background research, which helps identify key gaps in existing approaches and supports the formulation of the problem statement. After gathering insights from the preliminary research, the study refines the problem statement, pinpointing specific deficiencies in current RM for smart PSS, especially in relation to the integration of data-driven feedback and adaptive lifecycle considerations. A rigorous systematic literature review (SLR) is performed to systematically investigate the identified research gaps. After gathering the relevant literature, the research proceeds with an assessment of the findings with an emphasis on data processing. Each study is examined for insights into how it addresses RM for smart PSS design, particularly with regard to the challenges of data-driven feedback loops and cross-disciplinary collaboration.
The SLR was conducted in May 2024 to respond to the following research question: What methodologies exist for RM in the context of smart PSS development? Following a contextual literature review, a search protocol was devised (see Figure 1) and a search string was constructed for scientific databases. This search provides a comprehensive range of data, but it should be noted that it may also exclude other methods that are not included in these sources.

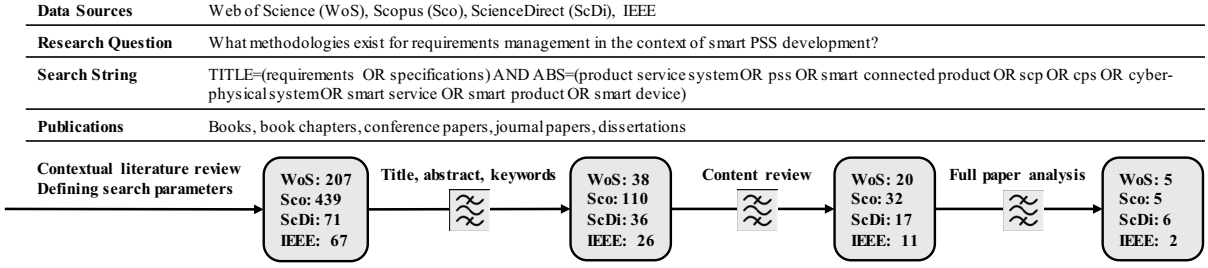


Figure 1. Search parameters used for the SLR

To gain a comprehensive understanding of RM in smart PSS development, adjacent areas such as service design and CPS were also included in the search. A search for approaches that are purely data-driven would be an ineffective use of resources in this context, as the existing literature assumes the presence of data-based methods in these areas. The utilized key-words and their synonyms were documented and constituted the foundation for the search string, which was then refined in an iterative process. The results were ultimately subjected to a three-stage review process. Stage one addressed the title, abstract, and keywords, stage two comprised a content review of relevant sections, and stage three featured a detailed analysis intended to ascertain whether the articles addressed specific RM methods. During the screening of the articles, the focus was on distinguishing between those that merely proposed a theoretical model and those that had already been applied to a real-life scenario. For the methodologies to be included in the final selection, they had to demonstrate a certain degree of fulfilment, allowing for their implementation in the development process.

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3. Findings

This chapter begins by presenting the identified methods. It then evaluates the suitability of RM methods for smart PSS development. Subsequently, the methods are analyzed within the context of data-driven design. Finally, principles for data-driven RM tailored to smart PSS development are proposed.

3.1. Description of identified methods for RM

The search identifies 16 distinct methods from different domains, such as PSS, CPS, or smart service design. Table 1 provides a brief description of their key elements for requirements management.

Table 1. Methods for requirements management

Borgne et al., 2016: The framework introduces FORM-L, a formal, model-based language for requirements engineering in smart industries. It is designed to reduce ambiguity and support simulation throughout a system's lifecycle. To facilitate the accessibility of FORM-L for domain engineers, the authors propose the graphical interface FORM-GL, which builds on FORM-L. This interface simplifies requirements modeling while enabling integration with tools like Modelica for simulation purposes. This approach facilitates early validation and the usability of formal methods in industrial environments.

Chen et al., 2020: The framework employs a rough-fuzzy best-worst method to identify and evaluate user activity-oriented service requirements for smart product-service systems. It maps user activities to smart capabilities, utilizing rough-fuzzy analysis to address intrapersonal and interpersonal uncertainties.

Han et al., 2017: The framework introduces the adapt-requirement model, which integrates goal models and problem frames for the analysis of requirements in self-adaptive CPS. This model combines stakeholder goals with system context specifications, thereby accommodating dynamic adaptation needs in CPS. Furthermore, it includes a UML profile and tool support for continuity with design processes.

Liu et al., 2019: The framework introduces an M-V-F model, integrating marketable forms, value systems, and key factors to co-generate requirements for smart PSS. Using a stakeholder value flow network, it captures requirements based on interdependencies among stakeholders. A revised rough-DEMATEL method then evaluates these requirements, accounting for the complex interactions and uncertainties within the PSS, to inform early-stage design decisions.

Liu et al., 2020: The framework employs a hybrid method comprising fuzzy DEMATEL, modified ANP, and fuzzy TOPSIS to analyze smart PSS. It focuses on customer needs, co-creative value propositions, and system requirements (SRs) to prioritize SRs and support effective design, addressing customer problems and facilitating mutual value creation.

Meng et al., 2021: The framework is centered on the identification and assessment of service requirements in smart product-service systems, with a particular emphasis on data-driven value creation. It employs a three-pronged approach to ascertain these requirements, considering the operational phases of the product, the types of data available, and the analytical technologies that can be leveraged. To evaluate these requirements, a hybrid method combining rough set theory and fuzzy logic is utilized, addressing the inherent uncertainties stemming from linguistic ambiguity and subjective biases.

Silva et al., 2023: The framework delineates the overarching requirements, with a particular emphasis on the optimization of the customer experience. By employing design thinking, the framework can distinguish the essential requirements of stakeholders. The framework focuses the implementation of features designed to facilitate real-time monitoring, enhance accessibility, and streamline maintenance.

Vuotto et al., 2019: The framework introduces REQV, a tool for automatic consistency checking of requirements in CPS. REQV employs property specification patterns to structure natural language requirements and translate them into formal logic. The tool identifies inconsistencies and provides a minimal set of conflicting requirements to assist users who do not have formal method-related expertise, with the objective of simplifying the verification and improving the accuracy of CPS requirement specifications.

Wang et al., 2019: The framework proposes a data-driven approach to requirements management for smart PSS, emphasizing the use of heterogeneous usage data and context-aware information. This four-layer architecture manages data collection, representation, reasoning, and evaluation to extract and dynamically prioritize user requirements. The approach enhances the accuracy of the identified requirements and reduces the influence of subjective bias.

Wang et al., 2021: This framework proposes a graph-based, context-aware approach for requirement elicitation in smart PSS. This method employs predefined ontologies and a graph embedding technique, Deepwalk, to model and analyze relationships between product, service, and context nodes. By leveraging user-generated and product-sensed data, the framework identifies implicit requirements.

(*Continued*)

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Table 1. Continued.

Wang et al., 2023: The framework presents a user-driven approach for the development of smart product-service systems. It combines rough set theory with logistic regression to identify key service elements based on user needs and to establish relationships between user perceptions and design features. By focusing on core service elements and stakeholder integration, the framework optimizes design for high user satisfaction.

Wiesner et al., 2017a: The framework addresses the engineering challenges inherent to the development of Cyber-Physical PSS (CPSS). It emphasizes the integration of physical and service components, while accounting for the inherent complexity resulting from the involvement of multiple stakeholders and disciplines. The key challenges identified include ensuring interoperability, facilitating stakeholder collaboration, and managing the evolution of the system over its lifecycle. The proposed approach combines requirements elicitation and management to support the development of CPSS across industrial applications.

Wiesner et al., 2017b: The framework examines the effects of environmental dynamism on requirements engineering for complex systems. It highlights the monitoring of changing stakeholder needs and technology to manage requirements over the system's lifecycle. The study proposes two approaches – the InnoScore Service and the Focus-Activity Model – to periodically assess and adapt requirements in response to environmental changes, ensuring alignment with evolving user and organizational contexts.

Zhang et al., 2023: The UNISON framework offers a systematic, data-driven approach to the elicitation and evaluation of user requirements in the context of smart PSS. The framework employs deep learning techniques, including Bi-LSTM and bi-term topic modeling, to analyze user-generated data from online reviews and extract both product- and service-related requirements. Sentiment analysis, the IPA-Kano model, and an opportunity algorithm are used to categorize and prioritize requirements, creating a landscape map for guiding improvements to smart PSS design.

Zhou et al., 2022: The framework is a user experience-oriented smart service requirement analysis model for the development of smart PSS. It comprises a four-phase process for the identification of user experience-oriented smart service requirements, integrating elements such as user context, activity journey, and smart capabilities. Furthermore, the framework introduces an asymmetric trapezium cloud-based best-worst method for the prioritization of these requirements in situations of uncertainty.

Zhou et al., 2023: The framework is centered on a smart experience-oriented customer requirement (SEO-CR) analysis for smart PSS, employing a novel hesitant fuzzy linguistic cloud DEMATEL method. This approach identifies and prioritizes SEO-CRs, considering uncertainties such as hesitancy, fuzziness, and randomness in the decision-making process.

3.2. Depiction of suitability of RM for smart PSS development

The success of a product, system, or service is fundamentally influenced by the quality of RM. Effective RM hinges on understanding user problems, translating them into precise requirements and product attributes, and generating viable solutions based on these insights. This process is shaped by various external factors and the interactions between stakeholders, which must be seamlessly integrated into the development workflows (Paliyenko et al., 2023). For smart PSS, this issue is further magnified by the need to define an increasing number of requirements for products, services, and interconnected systems within a multidisciplinary environment involving diverse stakeholders.

Traditional development approaches are often insufficient due to the dynamic, data-driven, and interdisciplinary nature of these systems (Paliyenko et al., 2024a). As such, RM for smart PSS must ensure that requirements remain relevant and adaptable across all lifecycle stages, from initial design to long-term operation and maintenance (Aurich et al., 2019). The ability to integrate lifecycle feedback into RM is a critical aspect of smart PSS development. These systems thrive on iterative improvement, where insights from later lifecycle stages, such as operational performance and user behavior, must feed back into the design and development phases. Traditional static methods often fail to accommodate this feedback, resulting in outdated or misaligned requirements (Paliyenko et al., 2024b; Zhou et al., 2022). Customer-centricity lies at the heart of smart PSS development, emphasizing the need to tailor solutions to individual user contexts. RM must account for the inherent context dependency of smart PSS, where the usage environment influences system behavior and value delivery. Continuous monitoring and adaptation to these contextual factors are essential for maintaining alignment with customer needs. Equally essential to smart PSS RM is the utilization of data as a core asset (Paliyenko et al., 2024b). Smart PSS generate vast amounts of data throughout their lifecycle, including usage logs, performance metrics, and maintenance records. Effective methods must not only capture this data but also translate it into actionable insights for refining requirements (Liu et al., 2020).

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The integration of product and service requirements represents another unique challenge in smart PSS development. Traditional methods often treat these domains separately, leading to inefficiencies and misalignments (Paliyenko et al., 2024a). In this context, managing the complexity of smart PSS is another critical consideration. As these systems grow in scale, they involve a broader range of stakeholders. RM must address the increasing volume of data and interdependencies. Hierarchical decomposition of requirements and modular design principles can simplify this complexity by breaking down requirements into manageable units and visualizing their relationships across the system. Traceability is particularly important in this context, enabling developers to track requirements from their initial high-level definitions through mid-level system specifications and low-level technical details. High-level requirements must be established prior to the development of the product or system. These requirements focus on the overarching business case and involve stakeholders from the value creation network. Mid-level requirements encompass system requirements and solution-specific requirements, addressing the smart PSS as a whole as well as its associated value creation processes. Low-level requirements pertain to more detailed aspects, including component-specific requirements, technical specifications, and service-related requirements.

In addition to addressing these core challenges, methods for smart PSS RM must align with data-driven design principles to support decision-making through evidence-based insights (Zhou et al., 2023). These methods should facilitate the identification of emerging trends and future user demands while employing data visualization to present clear and actionable insights to stakeholders. Additionally, project monitoring and control mechanisms should ensure that requirements consistently align with development goals, allowing for progress tracking and the early detection of potential deviations during the initial stages of the development process.

3.3. Assessment of methods for data-driven RM

The analysis of methods for RM reveals a wide range of approaches, each targeting specific aspects of the process. To gain a deeper understanding of how these frameworks integrate into data-driven engineering, they examined along four dimensions: data planning, data acquisition, data preparation, and application. These dimensions, derived from established frameworks like the CRISP-DM data cycle (Shearer, 2000) and ISO/IEC 8183 (2023) for AI system data processing, provide a structured lens for assessing the strengths and limitations of each approach. Notably, the dimension application was used in this paper instead of evaluation to emphasize the practical implementation of methods in case studies or use cases, rather than a formal evaluation based on predefined criteria. This distinction was made because the reviewed literature rarely conducted comprehensive evaluations along specific metrics. By focusing on applicability, this paper aims to provide a clearer understanding of the practical relevance and feasibility of the examined methods.

The analysis of the methods was conducted iteratively, initially by the authors, who specialize in systematic product and PSS development. In this process, each paper was examined for elements explicitly addressing data collection, analysis, preparation, and application within the context of RM. Particular attention was paid to the depth and clarity of the guiding methodologies presented, as the goal was to identify not just the presence of relevant elements but also the robustness and practicality of their implementation. This focus on methodological depth was essential to distinguish between superficial references to data-driven practices and genuinely actionable methods.

Following this initial analysis, the findings were reviewed and discussed with two additional academic researchers specializing in data-driven design and systems engineering. Their expertise provided valuable external perspectives, helping to refine the classification and interpretation of the methods. Insights gained from these discussions were incorporated into the paper.

The assessment of existing frameworks (see Table 2) shows that many focus predominantly on data collection and preparation, often to support the derivation of actionable requirements. However, the integration of these stages into a cohesive, lifecycle-oriented approach remains a key challenge for many methods. For instance, data planning is foundational for aligning the requirements management process with overarching project objectives and the broader system context. Despite its importance, detailed methodologies for executing data planning are often underdeveloped. Frameworks like Zhou et al. (2022) acknowledge the need for careful planning but lack concrete methods to achieve it, highlighting a significant gap in practical applicability.

Data acquisition, the systematic collection of information from various sources, is addressed with varying levels of detail. Some methods, such as those by Zhou et al. (2023), emphasize capturing customer emotions

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and experiences. While valuable for understanding user preferences, this focus often neglects other critical factors, such as environmental and system variables, which are essential for a holistic view of smart PSS requirements. On the other hand, frameworks like Wang et al. (2021) adopt innovative approaches, such as graph-based modeling, to contextualize requirements and explore the relationships between product and service components. This approach proves effective for mapping dependencies and supporting system-wide integration, although it may lack the granularity needed for certain technical requirements.

Data preparation, which involves cleaning, organizing, and transforming collected data for analysis, is consistently one of the strongest areas across most frameworks. Advanced methods leverage mathematical models to process complex datasets, enabling engineers to interpret and utilize data effectively. Frameworks such as those by Silva et al. (2023) and Zhang et al. (2023) excel in this dimension, integrating diverse data sources to inform decision-making processes. Silva et al. (2023), in particular, combine elements from existing development approaches to create a comprehensive framework validated through practical use cases, demonstrating its applicability in real-world scenarios. These efforts highlight the value of robust data preparation mechanisms in enhancing the effectiveness of requirements management.

Application, the final dimension, ensures that requirements align with stakeholder goals and system constraints throughout the lifecycle. This stage is crucial for validating the relevance and feasibility of requirements as they evolve. Frameworks like Zhou et al. (2022) and Wang et al. (2021) incorporate lifecycle-oriented evaluation strategies, enabling continuous adaptation of requirements in response to changing conditions. However, other approaches, such as those by Liu et al. (2019) and Wiesner et al. (2017a), remain limited to early-stage requirements elicitation and lack mechanisms for iterative validation, reducing their utility in dynamic smart PSS environments.

Table 2. Classification and assessment of identified RM methods

Method reference	Data planning	Data acquisition	Data preparation	Application of the method to a use-case
Borgne et al., 2016	1	1	3	N/S
Chen et al., 2020	1	3	3	Smart vehicle service system
Han et al., 2017	3	1	3	Smart firefighting system
Liu et al., 2019	1	2	2	N/S
Liu et al., 2020	3	3	3	Smart fridge
Meng et al., 2021	2	2	3	Smart vehicle
Silva et al., 2023	3	3	3	Smart airplane cabin
Vuotto et al., 2019	1	1	3	N/S
Wang et al., 2019	1	2	3	Smart living & manufacturing
Wang et al., 2021	1	2	3	Smart bike system
Wang et al., 2023	2	2	3	Service robot
Wiesner et al., 2017a	2	1	2	Smart maintenance
Wiesner et al., 2017b	1	2	2	N/S
Zhang et al., 2023	3	3	3	Smart cleaning robot
Zhou et al., 2022	1	2	2	Smart home
Zhou et al., 2023	1	2	3	Smart vehicle service system

Assessment of fitness: 3 = fully; 2 = only partly; 1 = rather not; N/S = not specified

A deeper exploration of the frameworks reveals that their focus on requirements types varies significantly. Some, like Liu et al. (2020) and Zhou et al. (2022), prioritize customer-centric requirements, emphasizing emotional and activity-based factors. These approaches align well with user-focused design principles but often underrepresent system-generated requirements derived from sensors or operational data. Conversely, frameworks by Meng et al. (2021) and Han et al. (2017) emphasize sensor-driven requirements, effectively capturing real-time insights from the product's operation. While both approaches offer valuable contributions, few frameworks successfully integrate customer-centric and system-centric requirements into a unified methodology.

Lifecycle integration is another critical factor influencing the suitability of these methods for smart PSS. Frameworks like Wang et al. (2021) and Zhou et al. (2022) stand out for their ability to support continuous requirements elicitation and adaptation across the product lifecycle. Their methods ensure that requirements remain responsive to evolving user needs and environmental conditions. In contrast,

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some frameworks, such as those by Wiesner et al. (2017a) and Borgne et al. (2016), focus heavily on specific development phases, limiting their applicability to iterative and adaptive lifecycle management. Overall, the analysis highlights that while many frameworks address specific aspects of requirements management effectively, significant gaps remain in their holistic applicability to smart PSS development. Data preparation is consistently well-addressed, with many methods providing robust tools for processing and interpreting data. However, data planning and acquisition require further development to ensure alignment with project goals and stakeholder needs. Similarly, while some methods excel in lifecycle integration, others remain limited to static or phase-specific approaches, reducing their effectiveness for dynamic smart PSS environments.

3.4. Principles for data-driven RM in smart PSS development

Drawing from the literature review and the analysis of existing methods in the context of data-driven design and lifecycle considerations, numerous valuable insights have been identified. This work derives essential principles to support the development of smart PSS and guide effective RM. To validate these proposed principles, they are carefully compared with the existing body of literature, which is referenced throughout this section.

The development of smart PSS requires advanced principles for data-driven RM to address the complexities of multidisciplinary collaboration, lifecycle integration, and the dynamic nature of user and environmental needs. These principles are essential to ensure that RM processes remain adaptive, effective, and capable of leveraging data as a strategic asset. By building on established theories in systems engineering, data management, and requirements elicitation, a comprehensive approach can be defined to guide RM practices in smart PSS contexts.

Data-driven RM must also *leverage the wealth of information* generated by smart PSS to inform decision-making and foster innovation. Systematic collection, analysis, and application of data throughout the requirements lifecycle allow RM processes to transition from reactive to proactive. Predictive analytics, natural language processing, and other advanced techniques can uncover latent requirements and anticipate future challenges and opportunities, enabling smart PSS to remain competitive and responsive. (Meng et al., 2021; Han et al., 2017)

A key principle of data-driven RM is its *alignment with the entire lifecycle* of smart PSS. Requirements must not only address the initial development phases but also evolve continuously during deployment, operation, and maintenance. This lifecycle-oriented approach ensures that insights gained from later stages, such as user behavior and system performance, feed back into early-phase decision-making. Adaptive RM processes enable smart PSS to respond to dynamic market conditions and evolving customer needs, ensuring their long-term relevance and effectiveness. Feedback loops, integrated into RM frameworks, capture data from usage monitoring and predictive maintenance to validate and refine requirements in real time, maintaining their applicability throughout the lifecycle. (Schweitzer, 2010)

The *quality and traceability of data* are critical to the success of data-driven RM. High-quality data that is accurate, complete, and representative of the system's operational context forms the foundation of effective decision-making. Traceability ensures that stakeholders can track the evolution of requirements from their initial definition to their implementation and eventual impact on system performance. (Paliyenko et al., 2022; Shearer, 2000)

Multidisciplinary collaboration is another essential principle, as smart PSS development involves diverse fields such as mechanical engineering, software development, and service design. Each discipline brings its own perspectives, goals, and terminologies, which can lead to misalignments if not managed effectively. RM methods must facilitate seamless communication and alignment among stakeholders. Tools like Model-Based Systems Engineering (MBSE) and smart platforms that provide tailored access to requirements can enable stakeholders to engage meaningfully with the RM process while preserving the integrity of interdisciplinary workflows. (Wang et al., 2019; Borgne et al., 2016)

Transparency in the RM process builds trust among stakeholders, fostering collaboration and reducing conflicts. Clear communication, accessible documentation, and accountability throughout the requirements lifecycle are key to maintaining this transparency. Practical implementations include version-controlled requirements repositories and stakeholder dashboards that provide real-time updates on the status and progress of requirements. By enhancing trust and decision-making, transparency strengthens stakeholder relationships and improves overall project outcomes.

Customer-centricity is a key objective of smart PSS, making it a central principle of data-driven RM. Requirements must be derived from a deep understanding of user needs, preferences, and pain points,

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ensuring the system delivers value throughout its lifecycle. Co-creation, usability testing, and customer feedback systems are viable techniques for achieving this. Such principles ensure that smart PSS remain aligned with user expectations, even as those expectations evolve over time. (Liu et al., 2020)

The context-dependent nature of smart PSS introduces additional complexity, requiring RM processes to be highly *adaptable and responsive* to external variables. Factors such as user preferences, geographical conditions, and regulatory requirements can significantly influence system behavior. Context-aware RM leverages data-based monitoring and contextual data analytics to capture and interpret these factors. Machine learning algorithms, for example, can identify patterns and trends in usage data, enabling the dynamic adjustment of requirements to reflect contextual realities. (Paliyenko et al., 2024b; Zhou et al., 2023)

The inherent complexity of smart PSS necessitates approaches that simplify and manage this complexity effectively. Modular RM frameworks, which break down complex systems into manageable components, can address subsystem-specific challenges without compromising the overall system. Modeling methods that map interdependencies between modules further enhance decision-making and system optimization, allowing stakeholders to manage the complexity of smart PSS with greater efficiency and clarity.

Scalability is another critical factor, particularly for large-scale smart PSS with extensive stakeholder networks and interconnected systems. Scalable RM methods ensure that processes remain effective as the system grows in size and complexity. Digital tools and cloud-based platforms provide the necessary infrastructure for managing large datasets, diverse requirements, and geographically distributed teams. This scalability ensures that RM practices can adapt to the dynamic demands of smart PSS development.

Continuous validation and verification are essential to ensure that requirements remain feasible and aligned with stakeholder goals. Integrated testing and validation processes, extending from initial requirements elicitation to post-deployment monitoring, are vital for maintaining alignment with real-world user expectation. Testing frameworks and simulation environments enable rapid validation, reducing the risk of misalignment or infeasibility. Such continuous validation processes ensure that smart PSS development stays on track and deliver systems that meet stakeholder expectations.

4. Discussion and outlook

The SLR reveals that there are already methods designed specifically for the RM of smart PSS. These methods effectively address isolated problems within RM and enable a more efficient development process. One important aspect to be highlighted is the integration into the development process of the vast amounts of data generated by smart PSS during their lifecycle: This is done to create value for both customers and companies. Many identified models address this issue using mathematical methods capable of processing large datasets. These methods support decision-making and help developers manage uncertainties. However, a persistent challenge in data and information processing lies in the lack of a unified descriptive logic that facilitates information exchange across the interdisciplinary domains of service, product, and software development.

The integration of lifecycle feedback into RM processes is another major focus area. Smart PSS operate across extended lifecycles, with requirements evolving in response to user interactions, environmental changes, and technological advancements. While the existing methods often target requirements elicitation and validation in the early stages, they fail to provide comprehensive approaches for managing the entire RM process. This is particularly the case when it comes to documenting and dynamically integrating new requirements throughout the lifecycle. Only two identified methods explicitly consider dynamic requirements and their integration into the smart PSS lifecycle. This limitation underlines the need for a holistic management approach that spans from requirements elicitation and validation to their dynamic integration and governance throughout the lifecycle. While many approaches excel in specific areas, such as data preparation or stakeholder involvement, few adequately address the holistic integration of lifecycle data, dynamic requirements evolution, and interdisciplinary alignment. For example, data planning, a foundational component of effective RM, is often underdeveloped, leaving practitioners without actionable strategies for aligning data collection with system objectives.

Leveraging lifecycle data enables RM processes to transition from reactive to proactive, allowing teams to anticipate and address emerging challenges before they escalate. Advanced analytics tools, such as predictive modeling and machine learning, play a critical role in this transformation by providing actionable insights that enhance decision-making and support innovation. Moreover, the principles emphasize data quality and traceability, ensuring decisions are based on reliable, transparent, and

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contextually relevant information. These principles address the growing complexity of smart PSS by enabling stakeholders to trace requirements across the system lifecycle, maintaining alignment with evolving objectives and performance metrics.

The multidisciplinary nature of smart PSS development introduces additional challenges to RM. Effective collaboration among mechanical engineers, software developers, and service designers requires RM methods that facilitate communication, alignment, and shared understanding. However, the lack of unified documentation practices in many existing approaches limits their effectiveness. The proposed principles advocate for MBSE techniques and collaborative platforms to bridge disciplinary divides and ensure a comprehensive, accessible documentation and communication process.

Context-awareness is another critical consideration, as smart PSS operate in dynamic and diverse environments. Requirements must not only reflect user preferences but also adapt to external factors such as regulatory constraints, environmental conditions, and market trends. The principles proposed in this paper emphasize the need for real-time monitoring and contextual data analytics to dynamically adapt requirements, ensuring alignment with the system's operational environment. However, while some methods support requirement elicitation during the lifecycle, they lack systematic approaches for integrating these requirements into the existing RM framework. Future work should explore strategies to address this gap, ensuring that dynamic requirements are incorporated into the broader RM process.

This paper provides a comprehensive investigation into existing frameworks and proposes principles for data-driven RM, addressing critical gaps in data planning, lifecycle integration, system-level requirement handling, and interdisciplinary collaboration. While this paper provides a robust foundation for data-driven RM in smart PSS, it is not without limitations. The proposed principles are derived from a synthesis of existing literature and theoretical insights, which may not fully capture the nuances of specific industrial applications. Practical validation of these principles through case studies or pilot projects is necessary to assess their real-world applicability and impact. Furthermore, the rapid pace of technological change poses an ongoing challenge for RM processes. As new tools, methods, and data sources emerge, RM frameworks must continuously evolve to remain relevant. This underscores the need for adaptive, iterative approaches that can incorporate advancements in data analytics, artificial intelligence, and other enabling technologies.

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