

Determining a Digital Engineering Framework: A Systematic Review of What and How to Digitalize

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This study is a systematic review to determine a conceptual framework for digital engineering, the objective being to select what and how to digitalize Department of Defense (DoD) acquisition processes, data, and decisions. The research question was, What are the best practices for Digitalization and Industry 4.0 to inform DoD acquisition programs? The study analyzed 20 peer-reviewed scholarly articles from the last five years, written by academics and practitioners from 19 countries, focused on Digitalization and Industry 4.0 methods and technologies. This study had five major findings: digitalization projects begin with strategic choices; digitalization is done within an ecosystem that constrains the technical options; digitalization requires a method of execution that assesses opportunity and limits risk; digitalization results in new processes using new data models that enable better decisions; feedback on that new business model will come internally from users and externally from customers.

Keywords: digital engineering, digitalization, Industry 4.0, framework, implementation, strategy

DETERMINING A DIGITAL ENGINEERING FRAMEWORK

DoD published its Digital Engineering (DE) Strategy in 2018. That was followed in 2020 by the Naval Digital Systems Engineering Transformation (DSET) Strategy. Both have the same five goals. The question has arisen of whether or not digital engineering (DE) is a new interdisciplinary branch of engineering, like systems engineering (SE) is a branch of industrial engineering. At this time, it has no distinct scientific principles applied to build particular things, no unique processes, methods or protocols; it is only a policy. However, the commercial world embraced Digitalization and Industry 4.0 out of necessity and has realized great opportunities that government can leverage.

Problem Statement

Executing acquisition plans in a predictable, fully resourced manner is challenging (Kraft, 2015). The National Defense Strategy states greater efficiency in procurement is a national priority (Department of Defense, 2018a). The National Defense Business Operations Plan declares that reforming the business processes is a key strategic goal (Department of Defense, 2018b). The resulting Digital Engineering Strategy admits that DoD lags industry on digital transformation solutions (Department of Defense, 2018c).

The DoD Digital Engineering Strategy has five Goals:

1. Formalize the development, integration and use of models to inform enterprise and program decision-making.
2. Provide enduring, authoritative source of truth.

3. Incorporate technological innovation to improve the engineering practice.
4. Establish a supporting infrastructure and environments to perform activities, collaborate, and communicate across stakeholders.
5. Transform the culture and workforce to adopt and support digital engineering across the lifecycle.

The DAU glossary defines digital engineering as “... *an integrated digital approach that uses authoritative sources of systems’ data and models as a continuum across disciplines to support life cycle activities from concept through disposal* (Defense Acquisition University, n.d.)” However, neither the goals nor the definition answers the critical questions of what or how to implement digitalization.

Rationale

Blackburn et al. (2018) was the report that formed the foundation of the DoD Digital Engineering Strategy, later restated and published in Bone et al. (2019). Neither articulated a conceptual framework for implementation. That is the rationale for this study.

Digital Engineering discussions often include unfamiliar and somewhat fluid terms. These may include Digital Thread (Kraft, 2020), Digital Twin (Madni et al., 2019), Digital Surrogate (Chakraborty et al., 2020), Electronic Prototype (Rieken et al., 2020), Authoritative Source of Truth (Kraft, 2019), Government Reference Architecture (Department of Defense, 2010), Open Architecture (Keller, 2021), and Agile Software (SAFe 5, n.d.). This study generally avoids them.

Objective

The objective of this study is to identify the current state of Digitalization practices and methods, and to identify a conceptual framework and notional integration of business processes to data products to structured decisions that would satisfy the goals of the DoD Digital Engineering Strategy. This study is a systematic review.

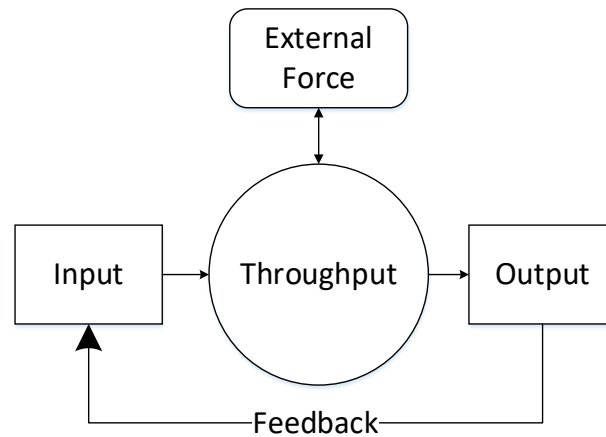
Potential Significance

Newly digitalized processes would be documented and constrained, with their triggers, inputs and outputs defined. Policy mandates imposed on a major defense acquisition program (MDAP) would be knowable and trackable over the life cycle of an acquisition program. Program decisions could be made with a common operating picture of the technical and managerial context around a given problem on a variety of levels, in a variety of functions, across the enterprise.

THEORETICAL FRAMEWORK

General Systems Theory (von Bertalanffy, 1972) provides a framework that can bridge between systems engineering, business process management, and decision science. A biologist, Bertalanffy published his Theory of Organic Shape, “Gestalt”, in 1926. He published his view of organisms as physical systems in 1940, and ultimately the seminal General Systems Theory (von Bertalanffy, 1950). A modern conceptual framework adapted from Marcketti et al. (2009) is shown in Figure 1. General Systems Theory Conceptual Framework.

FIGURE 1
GENERAL SYSTEMS THEORY CONCEPTUAL FRAMEWORK

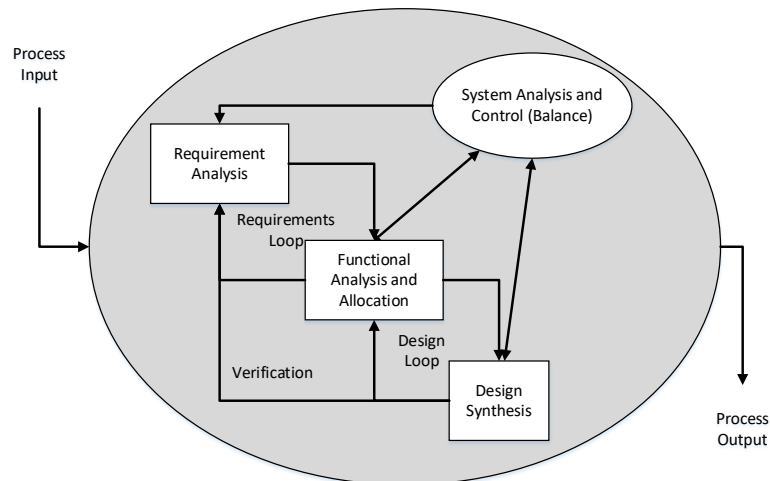


Note: Adapted from Marcketti et al. (2009)

Systems Engineering

The International Council on Systems Engineering (INCOSE) states that Systems Engineering emerged concurrently with Bertalanffy, at Bell Telephone Labs (INCOSE, n.d.). Hall (1962) defined a methodology for systems engineering to formalize and teach the principles of it. Kossiakoff & Sweet. (2002) cite several approaches, including one adopted by the Defense Acquisition University for instruction, shown in Figure 2. Systems Engineering Process. It bears note this engineering process is defined by inputs, a multi-step process, outputs, and feedback loops like Systems Theory.

FIGURE 2
SYSTEMS ENGINEERING PROCESS



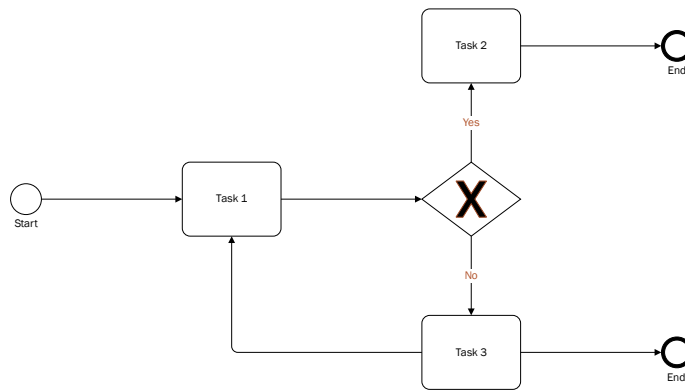
Note: Adapted from Defense Acquisition University (2001, p. 6)

Business Process Management

Dumas et al. (2013) state that Business Process Management (BPM) is how work should be performed in order to ensure consistent outputs and to take advantage of improvement opportunities. This includes a circular lifecycle of process identification, monitoring, modeling, analysis, and redesign. Business Process Model Notation (BPMN) is the industry standard and is defined by the Object Management Group (OMG),

as they do for Systems Modeling Language (SysML). Figure 3. Business Process Model and Notation Example shows the Microsoft Visio default process modeled with inputs, outputs, and feedback loop, also like Systems Theory.

**FIGURE 3
BUSINESS PROCESS MODEL AND NOTATION EXAMPLE**

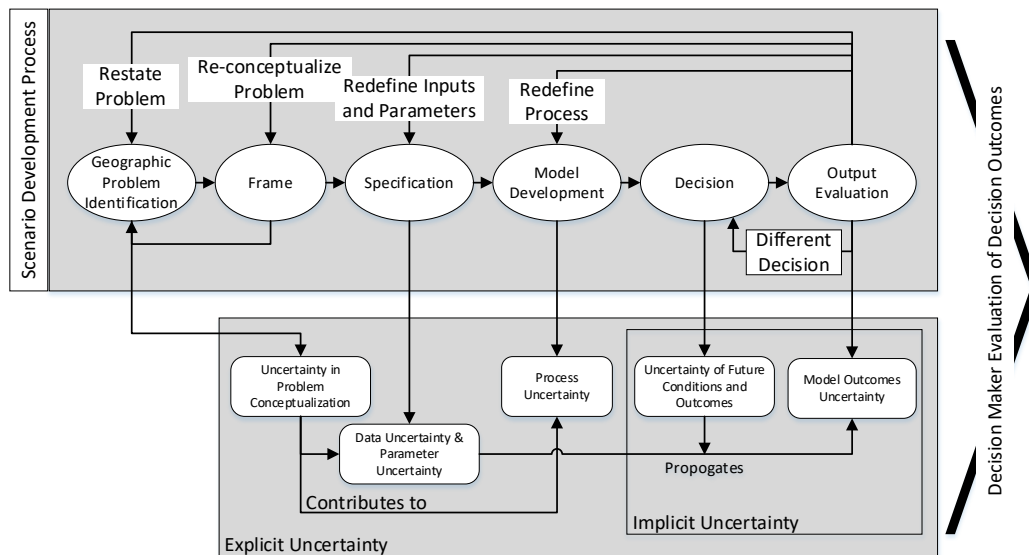


Note: Default example business process model in Microsoft Visio.

Decision Science

Davis et al., (2005) defined decision science (DS) as human decision making (why people decide) and the tools that assist it (decision support). Deitrick & Wentz (2015) discussed several theories in decision science. They showed that explicit and implicit uncertainty exist throughout the decision process, impacted in part by the changing interactions between steps in a process, the data, and the decision makers. It bears note they modeled a decision as a process with input, data, and outputs, as shown in Figure 4. Decision Process Diagram, again like Systems Theory.

**FIGURE 4
DECISION PROCESS DIAGRAM**



Note: Adapted from Deitrick & Wentz (2015, p.548).

Synthesis

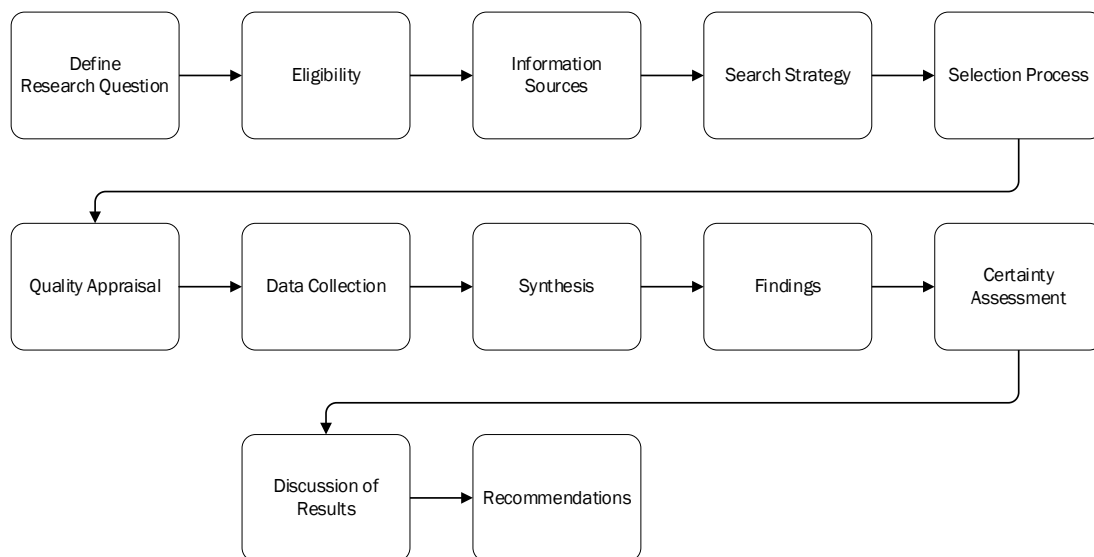
General Systems Theory (von Bertalanffy, 1972) is the parent theoretical lens for this study. It describes open systems as organisms that have input, throughput, output, react to external forces and have a feedback loop. Systems engineering as described by Kossiakoff & Sweet (2002) also has inputs, a process, output, feedback loops, and external forces. Deitrick & Wentz (2015) described decision processes with similar components. Inputs, output, feedback loops, and external forces are the archetypal objects in a BPMN process. Therefore, this framework is a convenient means to bridge these disciplines.

METHODOLOGY

General

This study is a systematic review of scholarly journals to provide the evidence-based current state of Digitalization and Industry 4.0 practice & methods. Petticrew & Roberts (2006) describe one method of performing a systematic review. Barends et al. (2017) offers a more streamlined approach for narrower questions that require rapid evidence assessments. The generally accepted method for more exhaustive review is the PRISMA, recently updated by Page et al. (2020). The PRISMA Checklist identified 27 items for consideration of inclusion in a systematic review. The research approach for this study was adapted from that, and is shown below in Figure 5. Overview of Research Approach.

**FIGURE 5
OVERVIEW OF RESEARCH APPROACH**



Note. Adapted from PRISMA Checklist (Page et al., 2020)

Research Question

The research question was developed using the PICOC (Population, Intervention, Comparison, Outcome, Context) framework (Barends et al., 2017). Initially, the population was to be the financial technology industry but scholarly research on that segment proved too narrow, so the population was broadened to general business. The intervention was Digitalization and Industry 4.0 practices & methods, as they are being applied by business operations. Comparison would be to the existing DoD practices. The outcome was improved business processes to create better data products to make better decisions. The context was DoD acquisition programs. Application of PICOC is summarized in Table 1 and led to development of the following research question, *What are the best practices for Digitalization and Industry 4.0 to inform DoD acquisition programs?*

**TABLE 1
PICOC FRAMEWORK**

Term	Description	Detail
Population	Who	Commercial Industry
Intervention	What or <u>How</u>	Digitalization and Industry 4.0 practices & methods
Comparison	Basis	Existing DoD processes, data products, and decisions
Outcome	Goal	Better planned and dynamic decision making, with lean processes.
Context	Circumstance	DoD acquisition program

Note: Adapted from Barends et al. (2017)

Eligibility

The inclusion criteria was restricted to peer-reviewed, scholarly journal articles. Only full-text articles were sought. For a fast-moving field, only articles from the last five years were accepted. While natural language processing translations provide extraordinary access, English language publications offer less risk of miscommunication. Sources were restricted to journals titled “business” or “management” that had published more than three articles on topic within the last three years, demonstrating sustained interest by the publisher, reviewers, authors and readers.

Information Sources

Google Scholar and ResearchGate were used to conduct initial scoping studies and find preliminary evidence on ‘digital engineering’ that address the current state of digitalization best practices, frameworks, strategies, and implementations, for process, or data, or decisions. The final search of UMGC OneSearch for evidence was reported using the PRISMA flow diagram (Moher et al., 2009).

Search Strategy

A set of Boolean search terms was developed with the assistance of UMGC librarians. Table 2 explains how they were derived. The final search set was: business AND technology, digitization OR digitalization, “best practice” OR framework, strateg* OR implement*, process OR data OR decision.

**TABLE 2
SEARCH TERMS & STRINGS**

Concept	Search Term
Technology Industry	business AND technology
Digitalization and Industry 4.0 practices & methods	digitization OR digitalization
Existing DoD processes, data products, and decisions	“best practice” OR framework AND strateg* OR implement*
	process OR data OR decision

Quality Appraisal Tools

For the study to be of value the source articles must be of quality. Critically appraising data sources prevents information overload, ensures relevance and is a best practice for evidence-based management (Rousseau, 2006). Weight of Evidence (Gough, 2007) was used to assess the coherence, appropriateness and relevance of articles. TAPUPAS (Pawson et al., 2003) was used to evaluate the selected articles for transparency, rigor, ethics, and quality assessments for inclusion.

Data Collection

Article meta data extraction involved collecting information such as year of publication, research design, sample size, population (e.g., industry, type of employees), and type of study. Overall

trustworthiness was judged. Core data extracted were the explicit findings, discussions, or conclusions of each article.

Synthesis

Collected data was recoded using the theoretical lens of Systems Theory, as relevant to the input, throughput, output, external force, or feedback of the open system. The categorized data was viewed for emergent themes. In the end, inputs clearly shaped strategic decisions, the throughput was the process of digitalization, external forces were part of the ecosystem or technical options, the output was a new business model, and feedback was provided by users and customers.

FINDINGS

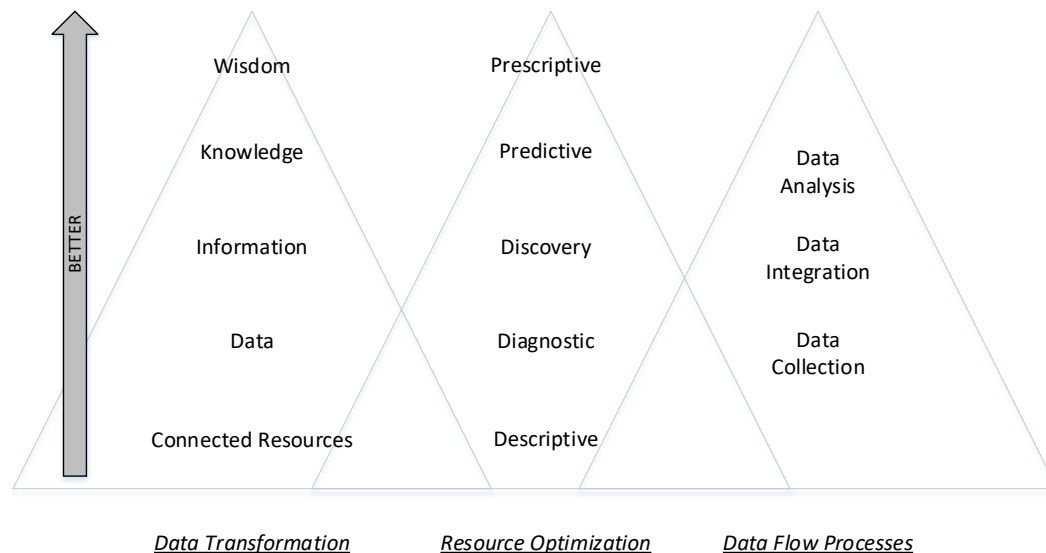
Input: Strategy Decisions

Blackburn et al. (2017) studied big data implications on research and development (R&D). They explored three important questions in the degree of change: how would big data refine, innovate, or transform R&D. Those mapped to impacts on strategy, people, technology and process.

Tortorella et al. (2021) explored the impact of Industry 4.0 on lean automation. They found process-oriented technologies had more impact on lean production than product and service technologies. This suggests a choice depending upon the desired target for impact.

Kristoffersen et al. (2020) proposed a Smart Circular Economy for manufacturing companies. This framework translates strategies into business analytics outcomes with digital technologies. It has three major dimensions that are relevant, each with degrees of implementation: Data Transformation, Resource Optimization, and Data Flow Process. This is shown in Figure 6.

**FIGURE 6
DECISION TOOL EXAMPLE**



Note: Adapted from Kristoffersen et al. (p248, 2020)

Nosalska et al. (2019) found Industry 4.0 to be a multidimensional system with numerous terms, categories and variables across its dualistic nature of technical and business. They documented the most common Industry 4.0 design principles over several years of publication. The top recurrent principles were flexibility, real-time capability, decentralization, and modularity.

There were warnings as well. Donnelly (2019) cautioned to avoid over-digitization, while encouraging formal and informal knowledge exchange. This was a key strategic consideration given the tensions of digital transformation.

Throughput: Process

Almost uniformly, the focus was not only on the process of how to digitalize, but emphasizing that business process is the most important target of digitalization. Specifically, the value of digitalization is realized through the transformed underlying business processes (Antonucci et al., 2021). Further, lean production is most affected by process technology (Tortorella et al., 2021). Lastly, process is a critical component of Industry 4.0 implementation in supply chains (Ghadge et al., 2020).

Janiesch et al. (2019) used the 6-step design science research (DSR) process for the design of autonomous agents in the Internet of Things (IoT). They described the DSR steps as 1. Problem Identification, 2. Objectives of a Solution, 3. Design and Development, 4. Demonstration, 5. Evaluation, and 6. Conclusion. They applied this to a scenario of a cyber-physical system (CPS), a self-driving car.

Linde et al., (2021) evaluated opportunities for digital modeling, and identified traps to avoid. They found a structured approach for evaluating digital business models had three phases: assessing the opportunity, managing risks, and modeling the future. Concurrently, Linde et al., (2021) found several common traps that must be avoided. First, companies in a rush may not understand the customer value they are creating, and fail to satisfy customer needs. Second, not understanding the value delivery process and how the new digitalized process fits within the rest of the corporate context has risks. Last, companies may not understand the new profit formula and means of realizing revenue, simply trusting that digitalization will have made things better.

Output: New Business Model

One critical component of Industry 4.0 implementation is the new digital business model (Ghadge et al., 2020). Another recurring theme is that technical and business-related aspects are interlocked factors (Nosalska et al., 2019). While business model change is enabled by digitalization (Laïfi & Josserand, 2016), the new business model progresses with the business modeling process (Mattsson & Andersson, 2019).

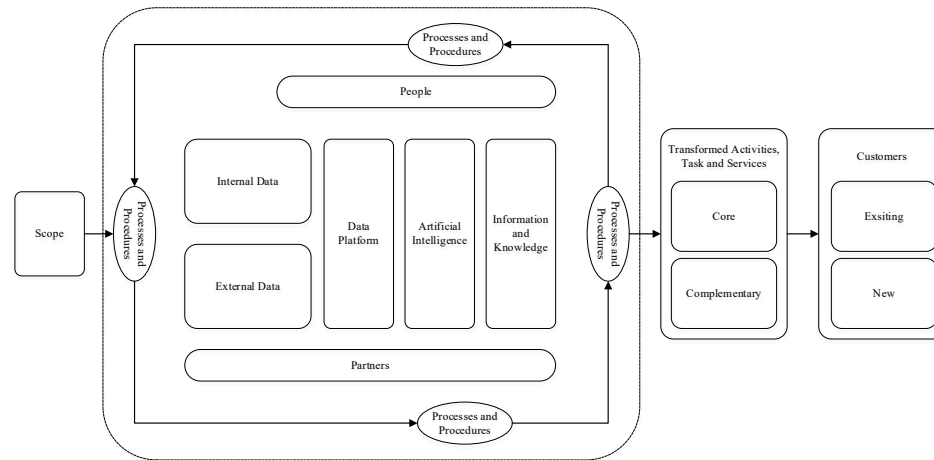
Particularly important for a government procurement agency, Mattsson & Andersson (2019) determined that public-private interaction reveals tensions that drive BPM: structural, behavioral, and organizational. Mattsson & Andersson (2019) concluded a public actor in the complex public network is a much more complex implementation.

External Force: Ecosystem and Technical

There are many external forces to consider. Cong et al. (2021) identified partners as part of the IoT ecosystem. Correani et al. (2020) described a digital transformation ecosystem in which the data platform worked with customers and other players. Dethine et al. (2020) suggested that the ecosystem adapts over time, as did Ghadge et al. (2020). Garay-Rondero et al. (2020) stated the ecosystem is digital and physical. Gastaldi et al. (2018) considered the firm's larger ecosystem important to a transformation. Linde et al. (2021) described the ecosystem in terms of relationships. Thus the ecosystem could be recoded as people, resources, organization and supply chain, and a digitalization project will have relationships with all of them.

Ivančić et al. (2019) identified seven main dimensions of digital transformation to include strategy, people, organization, customer, ecosystem, technology and innovation. In the framework shown below, Correani et al. (2020) included data sources, platform, and artificial intelligence (AI) as shown in Figure 7. Example Digital Implementation Framework. Ghadge et al. (2020) listed data sharing and management as critical.

FIGURE 7
EXAMPLE DIGITAL IMPLEMENTATION FRAMEWORK



Note: Adapted from Correani et al. (p45, 2020).

Nosalska et al. (2019) listed many Industry 4.0 key technologies such as CPS, Big Data, IIOT, Cloud Computing/ Cloud Manufacturing, Services/ Product-as-a-Service/ Internet of Services, and System/ Architecture. Tortorella et al., (2021) determined that some Industry 4.0 technologies are positively correlated with Lean Production (LP) practices, but not all. The emerging technical factors appear to be the platform, technologies, and the data.

Feedback: Users & Customers

Correani et al. (2020) stated that customers could give immediate feedback if part of the feedback loop. Donnelly (2019) noted opportunity to provide interpersonal feedback to clients and colleagues from digitalization. Garay-Rondero et al. (2020) found feedback to the value chain, and where physical processes affect computations and vice versa. Ghadge et al. (2020) found low customer service could be due to a backlog of feedback on demand. Mattsson & Andersson (2019) found small companies were quicker to adopt platform and content changes based on user feedback, and that customer feedback was important during development.

A Digital Community Infrastructure is digital sharing platforms to share designs and social networks and/or blogs to discuss ideas, questions and projects (Rieken et al., 2020). IoT technology and Artificial Intelligence of Things (AIoT) empowers the acceleration of digital transformation, and real-time collection of data from customers to monitor their conditions or assets to update risk. (Cong et al., 2021). Matzler et al. (2018) cautions that within the existing organization, implementation is highly unlikely to succeed, therefore organizational change is essential to success.

Certainty Assessment

Lewin et al. (2018) described a method of applying the Confidence in Evidence from Reviews of Qualitative research (CERQual) approach to identify the confidence in findings. CERQual is a framework to evaluate the methods, coherence, adequacy and relevance of the data used, effectively a self-report card that adds rigor and transparency.

DISCUSSION

Developing the Conceptual Framework

The proposed framework consists of input, throughput, output, feedback and external forces. In this model, the inputs are the strategic choices to be made for implementation:

1. Degree of Change (Blackburn et al., 2017).
Refine, or Innovate, or Transform.
2. Target for Lean Impact (Tortorella et al., 2021).
Process, or Product & Service.
3. Degree of Circular Economy (Kristoffersen et al., 2020).
Data Transformation, Resource Optimization, Data Flow Process.
4. Primary Design Principles: (Nosalska et al., 2019).
Flexibility, Real-Time Capability, Decentralization, Modularity.
5. Limit of Digitization (Donnelly, 2019).

These choices should be made with the intent of best achieving the DE Strategy goals of using models to inform decision making, creating an authoritative source of truth, technological innovation, supporting infrastructure and environments, and transforming the culture and workforce.

Throughput is the process of selecting processes, then digitalizing them. A best practice is to use 6-step design science research process (Janiesch et al., 2019). During execution, evaluate the opportunities and avoid the common traps (Linde et al., 2021). The method of digitally engineering is a process itself. That process is a mini-project plan for each business process under consideration of constraining the problem, setting goals, finding a solution, test, demonstration and deployment. Constraining the problem necessarily includes assessing the opportunity for process improvement, because some process improvements may not yield sufficient benefits to make the solutions cost effective or the margin for improvement may be too small. Lastly, as the process moves forward the team must continually assess risks. If the project understands the value the process creates (why we do it), the value delivery process (how we do it), and value realization (what we get out of it) those typical traps will be escaped.

The output is a new digitalized business model, where technical and business aspects are intertwined (Nosalska et al., 2019). The more the business model changes, the more the relationships with customers, the supply chain and internal users will change, and new opportunities will arise (Cong et al., 2021; Dethine et al., 2020; Garay-Rondero et al., 2020; Laïfi & Josserand, 2016). This the entire purpose of digitalization. While a DoD acquisition program does not realize revenue (they do not get 'paid' by the Pentagon for systems delivered), they certainly realize costs and deliver product. Having a well-documented business model, especially one that is digitally accessible will enable resource managers to see how their funds are being used, and will enable warfighters to see how their capabilities are being delivered. In addition, legislative authorizers and appropriators will be more easily persuaded to fund programs that are transparent to them.

Many external forces are at work, but they can be grouped into ecosystem constraints and technology opportunities. The ecosystem includes people, resources, organization and the supply chain, which the entity may or may not have control of (Cong et al., 2021; Correani et al., 2020; Dethine et al., 2020; Garay-Rondero et al., 2020; Gastaldi et al., 2018; Linde et al., 2021). Technical forces include the computing environment platforms, technologies, and data (Correani et al., 2020; Ghadge et al., 2020; Ivančić et al., 2019). Technologies do not equally benefit all desired outcomes (Tortorella et al., 2021), but several are key to Industry 4.0 application (Nosalska et al., 2019).

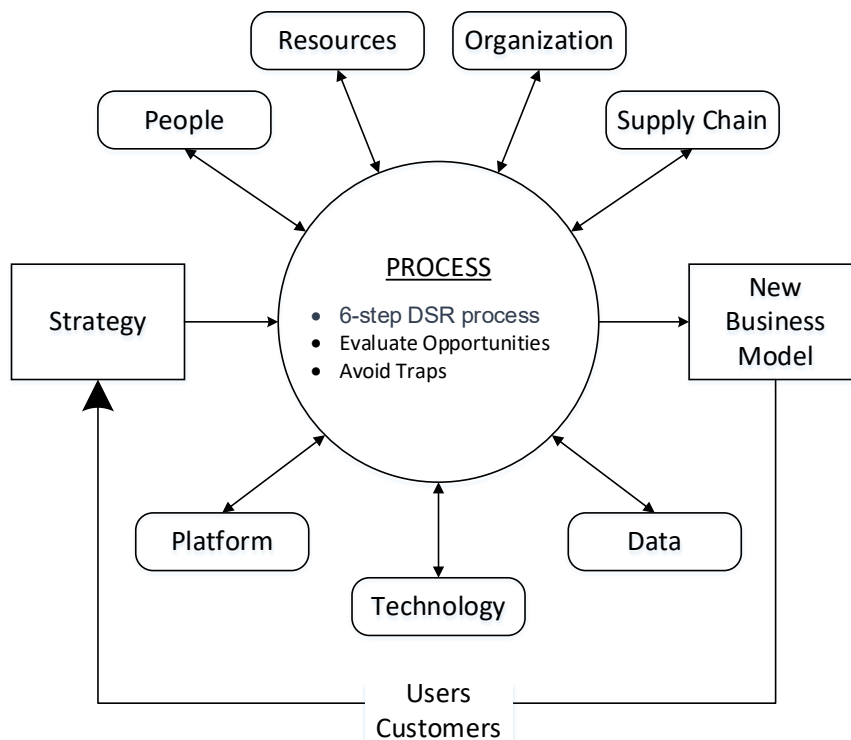
While the number of external forces at work could be infinite, the list must be constrained to provide meaningful decision points. The ecosystem forces were selected because their presence is necessary for success, even if they are constraints beyond the immediate control of the process owner. A process owner may not be able to change the people assigned, or may not have the authority to redirect resources, but both must be present in some limited quantity to succeed. A small operation may have complete control of its organization and culture, while many will be part of a larger organization with a set culture. Both can succeed, but the choices available are different. The digital supply chain for an office is crucial, and every office can identify who it depends on for data to execute owned processes, and what other offices consume data produced. Those players constitute the digital supply chain, and the participation of data suppliers and data consumers in digitally engineering a process is critical. The more they are integrated to the effort, the more opportunities may be exposed for further refinement, enhancing the recursive nature of digitalization.

Technical forces are more likely to be options than constraints. This is where people naturally gravitate to when considering digitalization. An office must consider its computing environment (platform), the technologies available (and affordable), and the data repositories it will require, create, and share. A small office may be able to change its platforms, whereas a larger office inside a large organization may have no control, or limited choices within a menu. A major choice will be between on-premises (e.g. desktop) and off-premises (e.g. cloud) computing, and that choice could be driven by security considerations. Industry 4.0 technologies are centered on IoT, and there are many technologies associated with that. Application of technologies like AI, ML, NFC or Bluetooth may accelerate IoT deployment, or they may have limited impact efficacy; being judicious is important. The new business model will hinge on the new data model. Businesses can collect data they never use, or fail to relate or visualize the data they have in a usable manner. A vast repository of stove piped data serves nobody. Data that is interrelated cross-functionally is more likely to have meaning. Data should be collected, created and shared because it is required to execute a process or make a decision.

Feedback will come first from internal users, eventually from external customers, as well as the digital supply chain (Cong et al., 2021; Garay-Rondero et al., 2020; Rieken et al., 2020). Communication with them is essential to success and subsequent adjustments. Providing a means for users to provide faster feedback via a Digital Community Infrastructure will lead to changes in the organizational culture and increase likelihood of acceptance, as users feel they are an integral part of changing the way they do their work. If feedback from those users is not aggressively sought, there is a risk they will obstruct change or sabotage the project. Those users must include not only the performers of a given process, but the users of its results, the decision makers. The best process with poor visualizations may not improve outcomes.

Figure 8 illustrates the derived conceptual framework for digital engineering. DoD Goals feed project strategy decisions, the ecosystem constrains technology choices, process defines execution, a new business model delivers efficiencies, and feedback informs recursion.

**FIGURE 8
DIGITAL ENGINEERING CONCEPTUAL FRAMEWORK**



The goal of digitalization is to arrive at a new set of processes that use a new set of data to achieve value. It is easy to see digitalization merely as a problem of new applications, or the introduction of Artificial Intelligence (AI) into processes, or new data models depending upon personal perspective or experience. However, none of those solutions alone will have sustained or meaningful impact. New models may be better, but may not result in better decisions if disconnected from a unified data model. A web services firm may be able to house petabytes of data for decades, but if it is not designed for people to use with their digital supply chain its customer value is limited. Using AI as support infrastructure to communicate with customers is common, but without integration with the business process it may not deliver value.

Entities have known they should digitalize, but did not know what or how to implement it. This framework provides a means to choose what projects to do and how to execute them in a balanced way.

RECOMMENDATIONS

Establish the implementation framework. Decide what external forces are strengths, weaknesses, opportunities or threats. To achieve DE goals, decide the strategy. Determine the desired degree of change, impact target, circular economy, design principles, and delimit the changeable processes. Model those processes as-is, to-be, and assess risk, as part of a disciplined project plan. Engineer a new data model on the proper platform with select technology, fed by new processes, and feeding others internally and externally. Communicate with the affected users, customers, and suppliers continuously, seeking failure early and rewarding good outcomes. Plan on necessary organizational changes. Monitor changes to the business model; prepare to adjust.

LIMITATIONS, IMPLICATIONS AND RISK

This systematic review was streamlined for rapid completion. While the search was conducted on UMGC library databases, a significant number of results were excluded based solely on the title or abstract, and may be subject to selection bias (Nunan et al., 2017). The search terms may be subject to selection bias. Article content may have been ignored or highlighted based on the authors experience, injecting confirmation bias (Spencer et al., 2018). Digitalization is a rapidly evolving practice with hotly competing providers who need a proprietary edge, which resists scholarly publication.

Engineering is commonly defined as the application of scientific principles to build things. The branches and sub-branches are differentiated by using particular scientific principles to build particular things: this is what differentiates mechanical engineering from software engineering. This paper associates the principles of systems engineering, business process management and decision science for the purposes of describing a DE framework.

The Accreditation Board for Engineering and Technology (ABET) certifies more than 3,000 programs at over 600 US institutions with 75 engineering programs, yet none are 'digital engineering.' Digital Engineering is not currently a defined branch of engineering; therefore, few journal articles reference it. DE might be a sub-branch sibling of Systems Engineering if a distinct DE process is proposed and accepted.

Entities have known they should digitalize, but did not know what or how to implement it. This framework provides a means to choose those projects and execute them in a balanced way. This framework is being deployed in a case study integrated product team (IPT) this summer.

According to Matzler et al. (2018) the biggest risk is the existing organization; therefore, companies need a new culture, with great incentives to innovate and small penalties for mistakes. Failing faster, cheaper, will lead to success in digital transformation.

CONCLUSION

Answer to the Research Question

The research question was: What are the best practices for Digitalization and Industry 4.0 to inform DoD acquisition programs? The study found broadly that an implementation framework is necessary to

properly apply Industry 4.0 technology to the digitalization of business processes. In the case of the DoD, the proposed framework shows Digital Engineering Strategy goals guide implementation decisions, the ecosystem constrains technology choices, an executable process is defined, the resulting new business model delivers efficiencies, and feedback informs recursion.

A conceptual framework was proposed that integrates these elements, as an evidence-based recommendation. A DoD agency that applied this method would be a cutting edge digitally engineered entity, capable of continuous digital evolution.

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ENDNOTE

- ¹ References marked with an asterisk (*) indicate studies included in the systematic review.

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