

Make Your Publications Visible.

A Service of



Leibniz-Informationszentrum Wirtschaft Leibniz Information Centre

Wohlleber, Annika; Münch, Christopher; Hartmann, Evi

Conference Paper

Prevailing technologies and adoption obstacles in maritime logistics

Provided in Cooperation with:

Hamburg University of Technology (TUHH), Institute of Business Logistics and General Management

Suggested Citation: Wohlleber, Annika; Münch, Christopher; Hartmann, Evi (2022): Prevailing technologies and adoption obstacles in maritime logistics, In: Kersten, Wolfgang Jahn, Carlos Blecker, Thorsten Ringle, Christian M. (Ed.): Changing Tides: The New Role of Resilience and Sustainability in Logistics and Supply Chain Management – Innovative Approaches for the Shift to a New Era. Proceedings of the Hamburg International Conference of Logistics (HICL), Vol. 33, ISBN 978-3-756541-95-9, epubli GmbH, Berlin, pp. 559-588, https://doi.org/10.15480/882.4717

This Version is available at: https://hdl.handle.net/10419/267199

Standard-Nutzungsbedingungen:

Die Dokumente auf EconStor dürfen zu eigenen wissenschaftlichen Zwecken und zum Privatgebrauch gespeichert und kopiert werden.

Sie dürfen die Dokumente nicht für öffentliche oder kommerzielle Zwecke vervielfältigen, öffentlich ausstellen, öffentlich zugänglich machen, vertreiben oder anderweitig nutzen.

Sofern die Verfasser die Dokumente unter Open-Content-Lizenzen (insbesondere CC-Lizenzen) zur Verfügung gestellt haben sollten, gelten abweichend von diesen Nutzungsbedingungen die in der dort genannten Lizenz gewährten Nutzungsrechte.



https://creativecommons.org/licenses/by-sa/4.0/

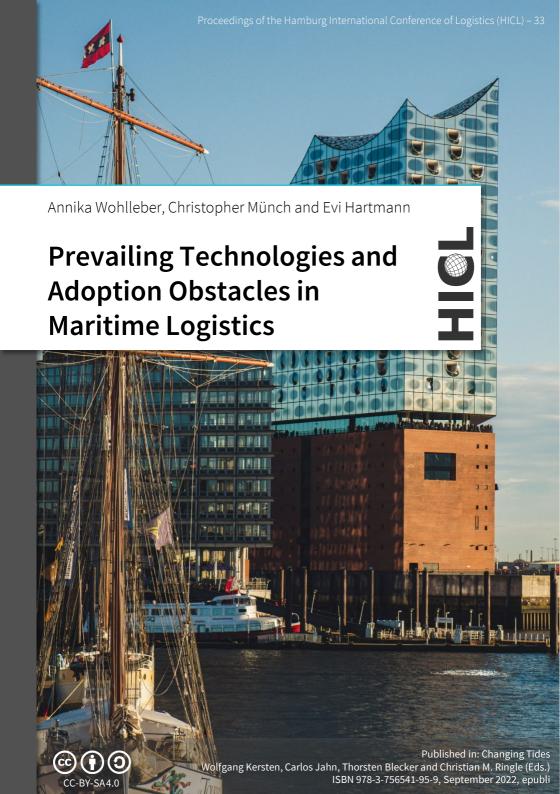
Terms of use:

Documents in EconStor may be saved and copied for your personal and scholarly purposes.

You are not to copy documents for public or commercial purposes, to exhibit the documents publicly, to make them publicly available on the internet, or to distribute or otherwise use the documents in public.

If the documents have been made available under an Open Content Licence (especially Creative Commons Licences), you may exercise further usage rights as specified in the indicated licence.





Annika Wohlleber¹, Christopher Münch¹ and Evi Hartmann¹

1 – Friedrich-Alexander-Universität Erlangen-Nürnberg

Purpose: Digital technologies (DTs) are transforming logistics operations in the shipping industry. Yet, the industry is only in the early stages of digitalization. Consequently, there is a lack of empirical evidence on applied DTs and related obstacles for DT adoption.

Methodology: A multiple case study was conducted comprising ports, freight forwarders, and carriers. Based on 18 expert interviews and additional data sources, differences and similarities concerning currently applied DTs and associated obstacles to DT adoption were examined.

Findings: Presented findings indicate comprehensive efforts toward a paperless and digitalized way of operating within maritime container logistics (MCL) by using DTs such as blockchain, cloud solutions, and artificial intelligence. Especially ports strive to achieve collaborative data usage but are frequently hindered by a lack of inter-organizational data sharing. Furthermore, the inherent complexity of the MCL chain and employees' and managers' defensiveness toward technological change needs to be overcome by applying suitable measures for DT adoption.

Originality: The research contributes to the scarce literature of DT adoption within MCL by providing empirical insights into the state-of-the-art of DTs for ports, freight forwarders, and carriers. Additionally, this research is the first to address implications for tackling existing obstacles for successful DT adoption in MCL.

First received: 17. Mar 2022 Revised: 12. Aug 2022 Accepted: 15. Aug 2022

1 Introduction

Maritime logistics is of outstanding importance for global supply chains, being responsible for the majority of overall freight logistics (OECD, 2022). In 2020, maritime container logistics (MCL) was in charge of the shipment of approximately 20% of tons loaded in overall international maritime logistics (UNCTAD, 2021). From 2022 to 2026, MCL is expected to grow 4.6% annually - thus exceeding the growth expectation of overall maritime logistics of 2.4% annually (UNCTAD, 2021), emphasizing its current and future significance for global supply chains. However, the container shortage, starting in 2020, negatively influenced the reliability and increased costs in MCL (UNCTAD, 2021), displaying the need to incorporate digital technologies (DTs) in MCL to compensate for the negative effects on global supply chains (Panwar, Pinkse and Marchi, 2022). Emerging DTs, such as artificial intelligence (AI), machine learning (ML), blockchain, or the Internet of things (IoT), are characterized by not merely digitizing products and processes of organizations, but beyond being responsible for the digitalization of organizations, radically modifying logistics chains, renewing business models, and affecting the logistics industry's structures (Ceipek, et al., 2021). Also before the COVID-19-crisis, the significance of emerging DTs in MCL was highlighted (Fruth and Teuteberg, 2017), as the following examples illustrate: (1) DTs, such as the Internet of Things (IoT), are applied to allow seamless tracking and tracing of containers (Sanchez-Gonzalez, et al., 2019); (2) blockchain technology is used for the standardization and digitization of paper-based processes (Yang, 2019); (3) sensors, among other things, support monitoring activities within ports (Fruth and Teuteberg, 2017); and (4) artificial intelligence (AI) was found to optimize routing problems (Jurdana, Krylov and Yamnenko, 2020) or to monitor and predict weather and ice conditions (Benz, Münch and Hartmann, 2021). These examples display that DTs have a huge potential for transforming MCL towards increased transparency and real-time information availability (Bathke, et al., 2022). However, Munim, et al. (2020) indicated that the mere existence of DTs is not efficient until it is widely adopted in overall MCL.

Regarding appropriate DT adoption, the research in MCL is still in its infancy (Fruth and Teuteberg, 2017; Tijan, et al., 2021a). Although the adoption of specific DTs such as

blockchain (Yang, 2019) or AI (Jurdana, Krylov and Yamnenko, 2020) have already been examined, no differences between incumbent actors in MCL were elaborated. However, as MCL is moving towards digitalization at different speeds in different domains (Sanchez-Gonzalez, et al., 2019), the behavior of organizations regarding DT adoption may differ (Fruth and Teuteberg, 2017). Additionally, emerging DT adoption was found to be one of the biggest challenges that organizations in logistics currently face (Mathauer and Hofmann, 2019; Karakas, Acar and Kucukaltan, 2021), and thus, several obstacles need to be overcome for successful DT adoption (Cichosz, Wallenburg and Knemeyer, 2020; Tijan, et al., 2021a; Yang, Fu and Zhang, 2021). These potential obstacles to DT adoption are currently lacking research in MCL. Referring to potential inherent differences between the main incumbent actors of MCL—ports, freight forwarders, and carriers (Talley and Ng, 2013), this research aims to answer the following research questions (RQs):

RQ1: What is the state-of-the-art of DT adoption comparing ports, freight forwarders, and carriers?

RQ2: What are the obstacles that need to be overcome for successful DT adoption for ports, freight forwarders, and carriers?

To answer the RQs, an empirical case study is conducted. Thereby, the study is structured as follows: After providing theoretical background information about the topic, the underlying methodology is explained. Next, a cross-case analysis is conducted, discussing the findings gathered from the case study. Lastly, managerial and theoretical implications, limitations, and paths for future research are displayed.

2 Theoretical background

The objective of this paper is to demonstrate the current DT adoption status and obstacles to DT adoption in MCL. Therefore, first, theoretical information regarding DT adoption is provided, followed by an analysis of obstacles to DT adoption in general.

2.1 Adopting digital technologies

DTs have significant impacts on different levels of logistics organizations: (1) within logistics organizations regarding a change in business models and processes; (2) between logistics organizations considering governance and relational configurations; and (3) at the level of the logistics industry regarding disruptions to the status quo and the emergence of new product or service providers (Wang and Sarkis, 2021). In particular, in MCL, DTs enable standardization, digitization, and easing of paperwork (Yang, 2019). Additionally, DTs may minimize the employee role in MCL organizations to that of a system observer, as DTs have the potential to self-optimize processes (Jurdana, Krylov and Yamnenko, 2020) and to take over error-prone tasks (Bălan, 2020). This results in a simplification of freight calculations, and a decrease in the costs of fuels and human resources (Bălan, 2020; Jurdana, Krylov and Yamnenko, 2020).

When adopting DTs, the study by Yang, Fu and Zhang (2021) stated that organizations go through several stages depending on their level of technological intelligence and supply chain collaboration. The authors' developed matrix can be adapted to MCL, as presented in Figure 1. The first axis, technological intelligence, is defined as the degree of intelligence to which DTs are adopted in MCL (Schoenherr and Swink, 2015). Little technological intelligence means that traditional DTs, such as enterprise resource planning (ERP), transportation management systems (TMS), application programming interfaces (API), robotics process automation (RPA), data collection and visualization, or data processing technologies, are used in MCL (Munim, et al., 2020; Núñez-Merino, et al., 2020; Yang, Fu and Zhang, 2021), thus representing digitized solutions that merely convert analog into digital information. On contrary, high technological intelligence implies that real-time data can be processed by applying smart sensors, and predictive analyses are applied for forecasting and real-time planning (Yang, Fu and Zhang, 2021), consequently being related to digitalization. High technological intelligence is represented by IoT, augmented and virtual reality (AR/VR), additive manufacturing, blockchain, cloud and edge computing (Sanchez-Gonzalez, et al., 2019), as well as AI and machine learning (ML) (Sanchez-Gonzalez, et al., 2019; Jurdana, Krylov and Yamnenko, 2020; Munim, et al., 2020). The second axis represents the level of collaboration between MCL organizations (Cloutier, Oktaei and Lehoux, 2020). By inter-organizational

application, DTs are considered to lead to intelligent supply chains and involve connections between different actors within supply chains (Núñez-Merino, et al., 2020). Low collaboration means that DTs are solely applied intra-organizational, whereby high collaboration refers to the application of DTs and data sharing across several organizations (Cloutier, Oktaei and Lehoux, 2020; Yang, Fu and Zhang, 2021).

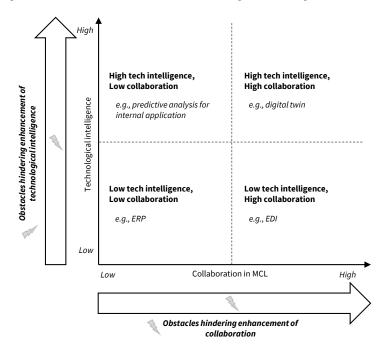


Figure 1: Research framework, adapted from Yang, Fu and Zhang (2021)

2.2 Obstacles to successful DT adoption in logistics

The mere existence of DTs has no impact until they are adopted in the whole MCL industry (Munim, et al., 2020). However, the inappropriate DT adoption potentially results in a disruptive change that leads to high risk and uncertainty (Yang, Fu and Zhang,

2021). Thus, DT adoption is referred to several potential disadvantages, such as cyber risks (Hofmann, et al., 2019), or the high potential volatility (Büyüközkan and Göçer, 2018). In addition, the uncertainty dilemma, stating that timely and accurate data is an arduous task and often not completely achievable in a volatile, uncertain, complex, and ambiguous environment, is a challenge regarding the usage of DTs (Lechler, Canzaniello and Hartmann, 2019). Consequently, obstacles inherent to DT adoption must be overcome to succeed in DT adoption (Yang, Fu and Zhang, 2021), as represented in the research framework in Figure 1. Potential general obstacles to DT adoption in logistics, found by screening respective literature, are summarized in Table 1. The results of Table 1 result from a desk research, using the databases Scopus and Google Scholar and applying the keyword sequence "digital technology" OR "digital technologies" AND "adoption" OR "implementation" AND "logistics" OR "supply chain" and additional relevant sources. The table serves as a basis for the analysis of the case study results regarding DT adoption obstacles in MCL.

Table 1: Obstacles to DT adoption according to literature

Obstacles to DT adoption	Reference(s)		
Heterogeneous information systems and lack of standards	Cichosz, Wallenburg and Knemeyer (2020), Inkinen, Helminen and Saarikoski (2019), Tijan, et al. (2021a), Yang (2019)		
Heterogeneous organizational structures or cultures	Harris, Wang and Wang (2015), Tijan, et al. (2021a)		
High implementation costs and risks	Harris, Wang and Wang (2015), Tijan, et al. (2021a)		
Lack of capabilities to change	Balci and Surucu-Balci (2021), Tijan, et al. (2021a)		

Obstacles to DT adoption	Reference(s)		
Lack of coordination and collaboration	Tijan, et al. (2021a), Yang (2019)		
Lack of early adopters	Balci and Surucu-Balci (2021)		
Lack of investments in DTs	Harris, Wang and Wang (2015), Tijan, et al. (2021a)		
Lack of knowledge about DTs	Balci and Surucu-Balci (2021), Tijan, et al. (2021a)		
Lack of regulation	Balci and Surucu-Balci (2021), Harris, Wang and Wang (2015), Tijan, et al. (2021a), Yang (2019)		
Lack of skills	Cichosz, Wallenburg and Knemeyer (2020),Tijan, et al. (2021a)		
Lack of support from stakeholders	Balci and Surucu-Balci (2021), Harris, Wang and Wang (2015)		
Lack of trust in DTs	Balci and Surucu-Balci (2021)		
Lack of urgency to adopt DTs	Tijan, et al. (2021a)		
Privacy concerns/no cyber security	Balci and Surucu-Balci (2021), Cichosz, Wallenburg and Knemeyer (2020), Harris, Wang and Wang (2015), Kala and Balakrishnan (2019), Tijan, et al. (2021a)		

Obstacles to DT adoption	Reference(s)
Resistance of stakeholders (e.g., employees) to adopt	Balci and Surucu-Balci (2021), Cichosz, Wallenburg and Knemeyer (2020), Tijan, et al. (2021a)

3 Methodology

The methodology applied to answer the RQs is a multiple-case design, collecting empirical data (Yin, 2018). Thereby, the following chapter is structured as follows: first, a short background about the case study design is provided. Afterward, the sample and data collection are presented in detail, followed by a description of the data handling.

3.1 Case study design

The case study approach is considered appropriate for our research for several reasons: (1) research regarding DT adoption in MCL is still at the beginning and existing knowledge regarding DT adoption is not sufficient (Fruth and Teuteberg, 2017); (2) the case study helps to widely evaluate the research problem and observe it within its actual practice (Yin, 2018); (3) the case study allows the close investigation of Yang, Fu and Zhang's (2021) DT adoption framework and expanding it based on the insights gathered from the interviews (Siggelkow, 2007); and (4) by using at least three sources of evidence per case, required triangulation is ensured (Eisenhardt and Graebner, 2007).

While executing the case study, construct validity, internal and external validity, and reliability need to be ensured to allow high-quality results (Yin, 2018). Construct validity is enabled by developing and adapting the questionnaire based on an extensive literature review and collecting multiple sources of data (Yin, 2018). Internal validity is allowed by analyzing existing literature regarding DT adoption and conducting interviews with a heterogeneous group of experts (Yin, 2018). External validity is ensured by conducting a multiple instead of a single case study; and reliability relies on the

selection based on predefined criteria, the sharing of the questionnaire in advance, and support by another researcher while analyzing the results (Yin, 2018).

3.2 Sample and data

We aimed to evaluate the differences between the three incumbent MCL actors "ports," "freight forwarders," and "carriers" (Talley and Ng, 2013). Thereby, the experts were selected according to the following, predefined criteria (Eisenhardt, 1989):

- Ports were required to be among the top 15 largest container ports around the
 world according to the handled volume. Experts from port authorities were
 chosen as interview partners in this category, as recent literature emphasized
 the significant role of port authorities regarding DT adoption in MCL,
 potentially becoming digital hubs in the future (Tijan, et al., 2021b);
- Freight forwarders needed to ship a volume of 500,000 twenty-foot-equivalent units per year by vessel; and
- Carriers were supposed to possess over 5% market share.

The literature provides several recommendations regarding the number of cases that seem appropriate for a multiple-case study. While some authors suggest that no more than 15 cases shall not be included (Perry, 1998), others propose that four to ten cases are sufficient (Eisenhardt, 1989). According to Corbin and Strauss (1990), additional cases shall be incorporated until saturation is achieved. They consider that any further cases would merely provide little variation compared to the already gathered data, marginally new insights, and no further relevant managerial and theoretical implications (Corbin and Strauss, 1990). Information regarding the final set of interview participants can be found in Table 2. The final set consists of nine cases and 18 interview participants. According to anonymization reasons, the size of the organizations is shown as an incremental range.

Table 2: Overview of cases

Case	Current position of the contact person	Organizational type	Number of employees	
Alpha ₁	Port Representative (Digital products)	Port authority	50,000- 100,000	
Alpha₂ Beta₁	Chief Digital & Innovation officer Project Manager for Digital Projects		200,000	
Beta ₂	Consultant for Digital and Business Transformation	Port authority	10,000– 50,000	
Gamma₁	Proposition Manager Digital Business Solutions	Port authority	>100,000	
Gamma₂	Product Lead Digital Products			
Delta ₁	Vice President and Global Head Ocean freight processes	Freight forwarder	>100,000	
Delta ₂	Head of International Supply Chain			
Epsilon₁	Vice President Global IT		50,000-	
Epsilon ₂	Senior Product Manager for Digital Innovation	Freight forwarder	100,000	
Zeta ₁	Vice President Global Sea freight Processes and Systems	Freight forwarder	50,000-	
Zeta₂	Vice President Global Sea Logistics Operations	Treignt forwarder	100,000	

Case	Current position of the contact person	Organizational type	Number of employees
Eta ₁	Member of the strategy office Digitalization Manager	Carrier	10,000– 50,000
Theta ₁	Deployment Lead (Transformation and Change Management)	Carrier	50,000- 100,000
Theta ₂	Investment Advisor Digital Products Global Chief Digital & Information Officer		50,000-
lota ₂	Projects Manager for Digital & Innovation	Carrier	100,000

3.3 Data handling

High construct validity was ensured by an extensive review of respective literature and discussions within the research team (Eisenhardt, 1989). This allowed the development of an interview guideline with a semi-structured design, using open questions, to react flexibly during the interviews (Yin, 2018). In addition, the interview questions were adapted during the conduction of the interviews. Before conducting the interviews, the potential experts were made familiar with the DT adoption framework by Yang, Fu and Zhang (2021) to understand the process of DT adoption. Furthermore, they were asked to classify their organization in this matrix.

All 18 interviews were conducted by two researchers of the research team. The interviewees had on average 14 years of experience in MCL. Data gathered from the semi-structured interviews served as the primary and most valuable data source (Yin, 2018). By additionally using secondary data besides the interview material, a potential social

desirability bias can be reduced (Crane, 1999). The interviews were then transcribed. Subsequently, the transcript and a summary of the interview were sent to each interviewee with the confirmation request to avoid misunderstandings and to guarantee the accuracy of the information supplied (Yin, 2018).

The data was analyzed by revealing common patterns and differences regarding DT adoption across the nine cases by a cross-case analysis. The identification of structures and patterns in the unstructured qualitative interview data and secondary data was allowed by a systematic coding process (Corbin and Strauss, 1990). Subsequently, categories were built on the identified codes. To allow inter-rater reliability and a high-qualitative analysis, an iterative coding approach was applied (Pagell and Krause, 2005). Consequently, another researcher of the research team revised and verified the identified codes and patterns during each coding step. Thereby, deviating results were iteratively approximated in three research meetings, leading to an agreement regarding the codes and patterns. Using the same categories for all interview transcripts allowed to efficiently compare the different case organizations (Eisenhardt, 1989). To facilitate the handling of the large amount of data, the categorization of the data was supported by computer-aided qualitative data analysis software (Yin, 2018).

4 Cross-case analysis

4.1 Digital technologies currently applied in maritime container logistics

According to the classification of the case organizations in Yang, Fu and Zhang's (2021) DT adoption matrix, Table 3 displays the level of technological intelligence and level of collaboration of the different case organizations.

Table 3: Current level of DT adoption of the nine cases

Case	Classification in DT adoption matrix	DTs applied in case organizations
Alpha	High tech intelligence/high collaboration	APIs, Autonomous shipping, Blockchain, Digital twin, Drones, ERP, IoT, Sensors
Beta	High tech intelligence/low collaboration	3D printing, Digital twin, Drones, ERP, ML, RPA, VR/AR
Gamma	High tech intelligence/low collaboration	AI, Autonomous shipping, Blockchain, Cloud computing, Drones, ERP, IoT, ML, RPA
Delta	Low tech intelligence/low collaboration	APIs, Chatbot, EDI, ERP, ML, Process mining
Epsilon	Low tech intelligence/high collaboration	APIs, AI, Chatbot, Cloud computing, Drones, ERP, IoT, ML, Quantum computing, RPA
Zeta	Low tech intelligence/high collaboration	APIs, AI, Blockchain, EDI, ERP, Predictive analytics, RPA
Eta	Medium tech intelligence/low collaboration	APIs, Blockchain, Cloud computing, ERP, RPA
Theta	Medium tech intelligence/medium collaboration	AI, Blockchain, Control tower solutions, ERP, IoT, ML, Process mining, RPA

Case	Classification in DT adoption matrix	DTs applied in case organizations
lota	Low tech intelligence/medium collaboration	APIs, AI, Blockchain, Cloud computing, ERP, IoT, Predictive analytics

All ports state that they hold a high technological intelligence. However, the level of collaboration differs among the interviewed experts. Whereas Alpha displays a high level of collaboration, Beta and Gamma do not show high collaboration efforts. Alpha strives toward establishing a "digital nervous system on top of the physical ports [...] to focus on safety and security in the port area "(Alpha₁), therefore testing the application of digital twins. This DT is also applied by Beta. However, Beta still displays silo-thinking and thus hinders collaboration efforts to some extent. All ports apply drones to "detect oil spills" (Alpha₁) or to "inspect buildings" (Beta₂). Blockchain is furthermore adopted by the ports for the transfer of documents. Additionally, Beta enhances its monitoring efforts by using sensors and AR.

Freight forwarders, on contrary, all display a low level of technological intelligence. While Delta further emphasizes its low level of collaboration, Zeta and Epsilon strive toward high collaboration with their MCL partners. Consequently, DTs to enhance data sharing are highly relevant for these two organizations. They adopted several cloud solutions to interact with their customers and manage their end-to-end logistics chain. Furthermore, Al simplifies their interaction with customers: "With Al, we have the repository, where we have a lot of replies to a lot of questions collected over time" (Epsilon₂) which can then be used for automated chatbots. Moreover, optical character recognition helps to scan physical documents that are arriving at the organizations: "We had a lot of business inquiries from potential customers: Can you fulfill this requirement? Or do you have this standard? [...] Now, we can do it in an hour instead, because we OCR scan these documents from the customers" (Epsilon₁).

The carriers are all located in the lower left-hand corner of the DT adoption matrix, having low to medium technological intelligence and low to medium collaboration. Two out of three carriers apply IoT in their organization. For example, Iota uses IoT for "the

monitoring of containers" and the establishment of smart containers. In addition, the application of AI, ML, or predictive analytics was emphasized by the experts. ML can help "analyzing all data that could eventually become [...] a new product that we could bring as part of our digital offering to our customers" (lota1). Some carriers strive towards establishing new services for their customers and therefore need to analyze data. Moreover, the experts mentioned that currently paper-heavy solutions are being replaced by digitized ones, such as electronic solutions for the bill of lading, and "many of the traditional paper-based architectures are or will be paperless" (Eta2). Tracking and tracing efforts by the application of blockchain and IoT are essential for the interviewed experts of the carriers. For Theta, it is not only relevant where the container is on the vessel, but beyond that "getting real-time visibility when your container is moving on a truck, where that truck is right now" (Theta1) to enable "new potential different products and services along that end-to-end journey" (Theta1).

The DTs adopted by the different case organizations are summarized in the following graph, presented in Figure 2.

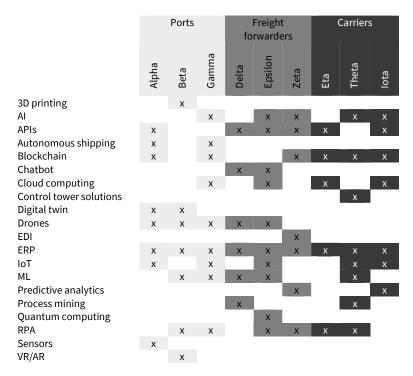


Figure 2: DTs adopted by different stakeholders in MCL

4.2 Obstacles to digital technology adoption in maritime container logistics

The obstacles to DT adoption in MCL can be classified into obstacles that hinder the enhancement of technological intelligence, obstacles that impede the enhancement of collaboration, as well as obstacles that hinder both. In this context, the obstacles mentioned during the interviews are displayed in Table 4. Table 1 in chapter 2.2 served as a basis for the analysis of the following results.

Table 4: Obstacles hindering DT adoption in MCL

No.	Obstacles to DT adoption	Examples	Barrier to technological intelligence	Barrier to collaboration
01	Competitors and external stakeholders Alpha, Delta, Eta	Market power of carriers; freight forwarders hindering DT adoption	No	Yes
02	Complexity of MCL network Alpha, Gamma, Delta, Epsilon, Zeta, Eta	Volatility of the market; involvement of many parties	Yes	Yes
03	Customers Alpha, Delta, Eta	Smaller customers; customers in specific countries	Yes	No
04	Heterogenous organizational structures Beta, Delta, Epsilon, Zeta, Theta, Iota	Size of the organizations; working in silos	Yes	No
О5	Heterogenous systems/lack of standards Alpha, Delta, Epsilon, Eta, Theta, lota	Plenty of different systems and interfaces; lack of standardization	No	Yes

No.	Obstacles to DT adoption	Examples	Barrier to technological intelligence	Barrier to collaboration
06	Internal bureaucracy Beta, Gamma, Delta, Epsilon, Zeta, Theta, Iota	Strong hierarchical structures; long decision processes	Yes	No
07	Lack of collaboration Alpha, Beta, Gamma, Delta, Zeta, Eta, Theta, Iota	No willingness to share data; lack of finding adequate partnerships	No	Yes
08	Lack of adequate government regulations Gamma, Delta, Epsilon, Zeta, Iota	No clear "digital" laws; plenty of different laws and regulations in different countries	Yes	Yes
09	Lack of necessary resources Alpha, Gamma, Delta, Epsilon, Eta, Theta, Iota	No employee capacity; lack of IT skills; lack of financial resources	Yes	No
010	Lack of strategy/urgency Alpha, Gamma, Eta, Theta, Iota	No clear vision; no shared business and technology strategy	Yes	Yes

No.	Obstacles to DT adoption	Examples	Barrier to technological intelligence	Barrier to collaboration
011	Lack of support from managers/C-level Beta, Gamma, Epsilon, Eta, Theta	Lack of management attention; lack of C- level attention	Yes	Yes
012	Lack of trust Alpha, Epsilon, Eta, Iota	No trust in systems or numbers; fear of losing competitive advantage	Yes	Yes
013	Legacy of old systems Alpha, Beta, Gamma, Delta, Epsilon, Zeta, Eta, Theta, Iota	Running costs for old systems; no innovation focus on internal processes	Yes	No
014	Old-fashioned industry Beta, Gamma, Delta, Epsilon, Zeta, Eta, Theta, lota	No job rotation; predominant incumbent organizations	Yes	Yes
015	Privacy concerns/no data security Alpha, Beta, Epsilon, lota	Fear of too much transparency of the organization; lack of data security	Yes	Yes

No.	Obstacles to DT adoption	Examples	Barrier to technological intelligence	Barrier to collaboration
016	Resistance of employees Beta, Gamma, Delta, Epsilon, Zeta, Eta, Iota	Older employees; employees hindering DT adoption	Yes	No

All experts mentioned that the legacy of old systems, such as "the amount of investments and support that we need to still do and have in regards of the legacy systems" (Zeta₁), hinders the investments and the available capacity for the adoption of emerging DTs. In this regard, the experts further emphasized that the maritime industry itself is old-fashioned with an antiquated culture and that "people in our industry generally are not quickly changing jobs. So, [...] our staff works at least 10 years in the company" (Eta₁). The missing collaboration is mentioned by 8 out of 9 organizations. However, especially within ports, many actors need to collaborate to guarantee efficient and disruption-free processes in the ports: "There is still a certain degree of competition among the port players and actually we should position ourselves more strongly as a closed port" (Beta₁).

Regarding the obstacle of the lacking resources, the organizations emphasize the challenge of "investments and a proper budget to be spent on something that we know that we may fail" (lota1). Besides "the competence of the workforce" (Gamma2), the challenge to attract the right "digital" skills to be able to cope with the respective DTs often hinders the adoption of emerging DTs. This is fueled by internal bureaucracy, as emphasized by all experts working in freight forwarding: "If something has to be researched, then you need a certain budget for it, then it has to be approved, and then [...] ten committees have to be passed" (Delta1). This is closely related to the resistance of employees. As several experts mentioned, especially older employees delay the DT adoption as they refuse to learn and adapt to new circumstances. Some employees fear being replaced by DTs. The employees further often lack trust in the systems and numbers: "They do not always trust the numbers, and they do not trust the Al results"

(lota₁), sticking to Excel sheets and thus hindering appropriate communication and analysis enabled DTs. This is related to privacy concerns between the different organizations, as mentioned by some experts: "Digitalization is one thing, but it has a flip side" (Alpha₂). Some organizations are worried about cyber risks resulting from emerging DTs.

In particular for carriers, both internal and external, heterogeneous systems and lacking standardization impede DT adoption. There is not "a unique platform, a unique standard that allows everyone to talk the same language when it comes to data" (Iota₁). The experts of freight forwarders emphasized the heterogeneous structures in MCL. Often, these organizations have "many siloed approaches between several teams" (Delta₁), leading to a strategy for DT adoption that is rather organized in silos. This is related to the complexity of the MCL network, as for "a container of food shipped from Africa to Europe, 30 parties [are] involved and 200 data transactions [are necessary]" (Alpha₁). Furthermore, several different laws and regulations exist on different continents and countries, being an obstacle to enhancing both collaboration and technological intelligence.

Additionally, the missing support by managers slows down DT adoption by being closely related to manager's priorities and alignment (Theta₂). In this context, several experts emphasized that a shared business and technology strategy is relevant to appropriate DT adoption: There is "a barrier between the business strategy and the technology strategy. So there are two different processes, [...] and there is always a bit of discrepancy that is a problem for implementing a common strategy and reinterpreting it with technology" (Theta₂). As further opined by several experts, some customers hinder the advancement of technological intelligence. "Many boxes continue to be booked by small customers. Small producers in China, small forwarders, and in many cases, they are still manual themselves" (Eta₁). Besides smaller customers in general that are less digitized, customers in specific regions impede DT adoption: "There are still a few small two-man companies in China that want to [communicate] by fax" (Delta₁). Competitors, in particular carriers, were identified as barriers by the other stakeholders as "there is a relatively small market of large carriers that cover the market and at the moment everyone is happy to get any capacity at all" (Delta₂).

The findings of the analyses in chapter 4.1 and chapter 4.2 are summarized in Figure 3, complementing the presented research framework.

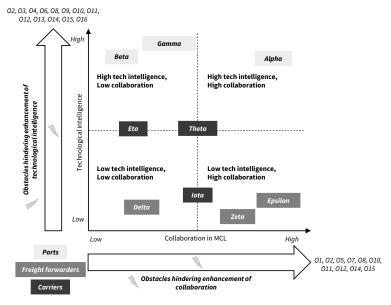


Figure 3: Expanded research framework, adapted according to Yang, Fu and Zhang (2021)

5 Discussion

Merely Alpha assessed themselves as being at a high level of both technological intelligence and collaboration. Beta, Gamma, Eta, Theta, Iota, and Delta need to overcome several obstacles hindering collaboration. As collaboration was found to have a positive impact on the competitive advantage of ports (Seo, Dinwoodie and Roe, 2016), also Beta and Gamma need to strive toward the upper-left-hand corner of the matrix. Zeta displays a high level of collaboration, but mentioned that it is difficult to find adequate partners for collaboration and ecosystems: "In South America, [...] Africa and

some areas in Asia, it is very difficult to find partners that would say let's see what we can do in terms of blockchain" (Zeta₂). The collaboration is hindered by trust, as one expert opined: "We make a decision only after we do all the steps that we used to do 15 years ago and a good feeling, [...], not looking in the numbers." (Eta₁) Furthermore, lacking data security is impeding the enhancement of trust: "We have massive problems driving digital projects forward because our lawyers and IT architects say, [...] before we do anything with the data, we have to make sure that it is protected, that it is secure." (Epsilon₁) In general, DTs, such as blockchain, were found to enhance trust by enabling cyber security, thus leading to increasing collaboration (Dubey, et al., 2020). Moreover, according to some experts, support from business associations, such as the digital container shipping association (DCSA), is necessary to develop and successfully implement standards regarding DT adoption. This is especially relevant due to the complexity of the MCL network (Jensen, Vatrapu and Bjørn-Andersen, 2018). Standardization is further emphasized regarding the "very different customs rules and customs processes in different countries" (Delta₂). Consequently, government support is essential. Nevertheless, in case blockchain is already adopted (Yang, 2019), standardization efforts can be advanced and thus other obstacles removed.

The obstacles hindering the enhancement of technological intelligence are mostly related to intra-organizational challenges. Maritime organizations mainly consist of old-fashioned people: "So, [...] our staff works at least 10 years in the company" (Eta1), which impedes bringing new insights regarding the usefulness of several DTs in the organization (Yang, 2019) because "traditional values are still predominant" (Beta2). As "there is a lot of demand for digital experts in various areas" (Epsilon2), MCL is having a problem attracting the right skills (Canepa, et al., 2021), for example in "software development and hardware engineering talent, to get enough people" (Theta2). The commitment of the management- and C-level is relevant for DT adoption (Ko, et al., 2022). As one expert aptly opined, "Our CEO once said, we are not Google. We are not an internet company. He is right, we are not, but the future is" (Eta2). This requires an adequate strategy for DT adoption. As mentioned by several experts, they often have a strategy regarding DT adoption, but they are "struggling with making the strategy a bit more tangible for the various subdivisions in the company and cascading it down to

management in a coherent way" (Theta₁). Thus, the strategy of DT adoption should be shared across organizational units, as there is an "enormous divergence in maturity, in understanding what digital is all about" (Alpha₁), dissolving silo approaches in the maritime organizations.

6 Conclusion, implications, and further research

As far as known by the authors, this research is the first to address the state-of-the-art of DT adoption and obstacles to further DT adoption in MCL by elaborating on similarities and differences between ports, freight forwarders, and carriers.

By applying a multiple-case approach, we identified that ports already display high levels of technological intelligence, but that collaboration efforts can be enhanced between the ports and external stakeholders with the support of governments and business associations such as the DCSA. DT adoption in ports is currently hindered by a legacy of old systems and ways of working within their organizations. The freight forwarders all display low levels of technological intelligence, but their collaboration level differs. They strive toward the application of DTs but are hindered by their silo approach. Their DT adoption is further impeded by the market power of the carriers. The carriers, however, are located in the midfield of the DT adoption matrix. Therefore, for carriers, obstacles for both leveraging the level of collaboration, such as heterogeneous systems, and the level of technological intelligence, such as their old-fashioned ways of working, need to be overcome. To summarize, our developed framework helps MCL organizations to derive implications for tackling the existing obstacles for successful DT adoption and to adapt their DT adoption strategy accordingly.

Theoretically, our research contributes by providing a detailed DT adoption framework for MCL, displaying the levels of DT adoption of the different incumbent maritime organizations based on the research of Yang, Fu and Zhang (2021). Our adapted framework highlights the obstacles that need to be overcome to further enhance the level of DT adoption in maritime organizations. Thereby, our research is the first to empirically provide such a framework in the context of MCL and supports incumbent maritime organizations in benchmarking themselves with other organizations.

The research exposes certain limitations that, in turn, reveal potential avenues for further research. One limitation regards the selection of the cases. Even though the number of cases seems sufficient according to Eisenhardt (1989), more research including smaller stakeholders in MCL may help to enhance the generalizability of the results, referring to required standardization efforts of the whole MCL industry. Additionally, merely organizations in MCL are considered for the research. In this context, further research needs to elaborate on whether the findings of this research are also applicable to other types of maritime logistics. Lastly, as MCL is inherently multimodal, especially the collaboration obstacles need to be regarded from an angle of collaboration with actors outside of MCL, incorporating other modes of freight transportation.

References

- Bălan, C., 2020. The disruptive impact of future advanced ICTs on maritime transport: a systematic review. *Supply Chain Management: An International Journal*, 25(2), pp. 157–175.
- Balci, G. and Surucu-Balci, E., 2021. Blockchain adoption in the maritime supply chain: Examining barriers and salient stakeholders in containerized international trade. *Transportation Research Part E: Logistics and Transportation Review,* 156, pp. 1–17.
- Bathke, H., Munch, C., Gracht, H. A. von der and Hartmann, E., 2022. Building Resilience Through Foresight: The Case of Maritime Container Shipping Firms. *IEEE Transactions on Engineering Management*, pp. 1–23.
- Benz, L., Münch, C. and Hartmann, E., 2021. Development of a search and rescue framework for maritime freight shipping in the Arctic. *Transportation Research Part A: Policy and Practice*, 152, pp. 54–69.
- Büyüközkan, G. and Göçer, F., 2018. Digital Supply Chain: Literature review and a proposed framework for future research. *Computers in Industry*, pp. 157–177.
- Canepa, M., Ballini, F., Dalaklis, D. and Vakili, S., 2021. Assessing the Effectiveness of Cybersecurity Training and Awareness Raising within the Maritime Logistics Domain. In: L. Gómez Chova, A. López Martínez, and I. Candel Torres. *INTED2021 Proceedings. 15th International Technology, Education and Development Conference*. Online Conference, 3/8/2021 3/9/2021: IATED, pp. 3489–3499.
- Ceipek, R., Hautz, J., Petruzzelli, A. M., Massis, A. de and Matzler, K., 2021. A motivation and ability perspective on engagement in emerging digital technologies: The case of Internet of Things solutions. *Long Range Planning*, 54(5), pp. 1–24.
- Cichosz, M., Wallenburg, C. M. and Knemeyer, A. M., 2020. Digital transformation at logistics service providers: barriers, success factors and leading practices. *The International Journal of Logistics Management*, 31(2), pp. 209–238.

- Cloutier, C., Oktaei, P. and Lehoux, N., 2020. Collaborative mechanisms for sustainabilityoriented supply chain initiatives: state of the art, role assessment and research opportunities. *International Journal of Production Research*, 58(19), pp. 5836– 5850.
- Corbin, J. M. and Strauss, A., 1990. Grounded theory research: Procedures, canons, and evaluative criteria. *Qualitative Sociology*, 13(1), pp. 3–21.
- Crane, A., 1999. Are You Ethical? Please Tick Yes □ Or No □ On Researching Ethics in Business Organizations. *Journal of Business Ethics*, 20(3), pp. 237–248.
- Dubey, R., Gunasekaran, A., Bryde, D. J., Dwivedi, Y. K. and Papadopoulos, T., 2020. Blockchain technology for enhancing swift-trust, collaboration and resilience within a humanitarian supply chain setting. *International Journal of Production Research*, 58(11), pp. 3381–3398.
- Eisenhardt, K. M., 1989. Building Theories from Case Study Research. *Academy of Management Review*, 14(4), pp. 532–550.
- Eisenhardt, K. M. and Graebner, M. E., 2007. Theory Building From Cases: Opportunities And Challenges. *Academy of Management Journal*, 50(1), pp. 25–32.
- Fruth, M. and Teuteberg, F., 2017. Digitization in maritime logistics—What is there and what is missing? *Cogent Business & Management*, 4(1), pp. 1–40.
- Harris, I., Wang, Y. and Wang, H., 2015. ICT in multimodal transport and technological trends: Unleashing potential for the future. *International Journal of Production Economics*, 159, pp. 88–103.
- Hofmann, E., Sternberg, H., Chen, H., Pflaum, A. and Prockl, G., 2019. Supply chain management and Industry 4.0: conducting research in the digital age. *International Journal of Physical Distribution & Logistics Management*, 49(10), pp. 945–955.
- Inkinen, T., Helminen, R. and Saarikoski, J., 2019. Port Digitalization with Open Data: Challenges, Opportunities, and Integrations. *Journal of Open Innovation: Technology, Market, and Complexity*, 5(30), pp. 1–16.

- Jensen, T., Vatrapu, R. and Bjørn-Andersen, N., 2018. Avocados crossing borders: The problem of runaway objects and the solution of a shipping information pipeline for improving international trade. *Information Systems Journal*, 28(2), pp. 408– 438.
- Jurdana, I., Krylov, A. and Yamnenko, J., 2020. Use of Artificial Intelligence as a Problem Solution for Maritime Transport. *Journal of Marine Science and Engineering*, 8(3), pp. 1–9.
- Kala, N. and Balakrishnan, M., 2019. Cyber Preparedness in Maritime Industry. International Journal of Scientific and Technical Advancements, (5), pp. 19–28.
- Karakas, S., Acar, A. Z. and Kucukaltan, B., 2021. Blockchain adoption in logistics and supply chain: a literature review and research agenda. *International Journal of Production Research*, pp. 1–24.
- Ko, A., Fehér, P., Kovacs, T., Mitev, A. and Szabó, Z., 2022. Influencing factors of digital transformation: management or IT is the driving force? *International Journal of Innovation Science*, 14(1), pp. 1–20.
- Lechler, S., Canzaniello, A. and Hartmann, E., 2019. Assessment sharing intra-industry strategic alliances: Effects on sustainable supplier management within multitier supply chains. *International Journal of Production Economics*, 217, pp. 64–77.
- Mathauer, M. and Hofmann, E., 2019. Technology adoption by logistics service providers. International Journal of Physical Distribution & Logistics Management, 49(4), pp. 416–434.
- Munim, Z. H., Dushenko, M., Jimenez, V. J., Shakil, M. H. and Imset, M., 2020. Big data and artificial intelligence in the maritime industry: a bibliometric review and future research directions. *Maritime Policy & Management*, 47(5), pp. 577–597.
- Núñez-Merino, M., Maqueira-Marín, J. M., Moyano-Fuentes, J. and Martínez-Jurado, P. J., 2020. Information and digital technologies of Industry 4.0 and Lean supply chain management: a systematic literature review. *International Journal of Production Research*, 58(16), pp. 5034–5061.

- Organisation for Economic Co-Operation and Development (OECD), 2022. Ocean shipping and shipbuilding. [online] Available at: https://www.oecd.org/ocean/topics/ocean-shipping/ [Accessed 7 March 2022].
- Pagell, M. and Krause, D. R., 2005. Determining when multiple respondents are needed in supply chain management research: the case of purchasing and operations. *Academy of Management Proceedings*, 2005(1), 1-5.
- Panwar, R., Pinkse, J. and Marchi, V. de, 2022. The Future of Global Supply Chains in a Post-COVID-19 World. *California Management Review*, 64(2), pp. 5–23.
- Perry, C., 1998. Processes of a case study methodology for postgraduate research in marketing. *European Journal of Marketing*, 32(9/10), pp. 785–802.
- Sanchez-Gonzalez, P.-L., Díaz-Gutiérrez, D., Leo, T. J. and Núñez-Rivas, L. R., 2019. Toward Digitalization of Maritime Transport? *Sensors (Basel, Switzerland),* 19(4), pp. 1–22.
- Schoenherr, T. and Swink, M., 2015. The Roles of Supply Chain Intelligence and Adaptability in New Product Launch Success. *Decision Sciences*, 46(5), pp. 901–936.
- Seo, Y.-J., Dinwoodie, J. and Roe, M., 2016. The influence of supply chain collaboration on collaborative advantage and port performance in maritime logistics. *International Journal of Logistics Research and Applications*, 19(6), pp. 562–582.
- Siggelkow, N., 2007. Persuasion With Case Studies. *Academy of Management Journal*, 50(1), pp. 20–24.
- Talley, W. K. and Ng, M., 2013. Maritime transport chain choice by carriers, ports and shippers. *International Journal of Production Economics*, 142(2), pp. 311–316.
- Tijan, E., Jović, M., Aksentijević, S. and Pucihar, A., 2021a. Digital transformation in the maritime transport sector. *Technological Forecasting and Social Change,* 170, pp. 1–15.

- Tijan, E., Jović, M., Panjako, A. and Žgaljić, D., 2021b. The Role of Port Authority in Port Governance and Port Community System Implementation. *Sustainability*, 13(5), p. 2795–2795.
- United Nations Conference on Trade and Development (UNCTAD), 2021. Review of Maritime Transport 2021. https://unctad.org/system/files/official-document/rmt2021_en_0.pdf [Accessed 23 June 2022].
- Wang, Y. and Sarkis, J., 2021. Emerging digitalisation technologies in freight transport and logistics: Current trends and future directions. *Transportation Research Part E: Logistics and Transportation Review*, 148, pp. 1–8.
- Yang, C.-S., 2019. Maritime shipping digitalization: Blockchain-based technology applications, future improvements, and intention to use. *Transportation Research Part E: Logistics and Transportation Review*, 131, pp. 108–117.
- Yang, M., Fu, M. and Zhang, Z., 2021. The adoption of digital technologies in supply chains: Drivers, process and impact. *Technological Forecasting and Social Change*, 169, pp. 1–13.
- Yin, R. K., 2018. Case study research and applications: Design and methods. Los Angeles, London, New Dehli, Singapore, Washington DC, Melbourne: Sage.