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RESEARCH ARTICLE

Adoption of Cybersecurity in the Chilean Manufacturing Sector: A First Analytical Proposal

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ABSTRACT This paper focuses on adopting cybersecurity procedures in Chilean manufacturing companies in the context of the Fourth Industrial Revolution's data-driven demands, which have exposed vulnerabilities in cybersecurity. This analysis is based on data from the Fifth Longitudinal Survey of Companies - ELE 5 - conducted by the National Institute of Statistics. Using the TOE adoption model, we employ binary and ordered Logit and Probit models with data from 574 companies and 17 explanatory variables. The objective is to gain insight into the factors influencing the adoption of cybersecurity processes, complementing the existing literature, which often focuses on developing specific technologies or conducting comprehensive analyses of digital transformation. The study highlights the significance of company size in explaining the adoption of cybersecurity procedures and reveals the relevance of explanatory variables as the depth of adoption increases. The findings underscore the need for public policies that facilitate the implementation of existing regulations, such as ISO 27.001, particularly for small companies. Additionally, the study emphasizes the importance of fostering a "culture of cybersecurity" across different sectors of society.

INDEX TERMS Adoption, cybersecurity, factory, industry 4.0.

I. INTRODUCTION

The Fourth Industrial Revolution, prominently features the implementation of sensors in processes and products, the use of cloud computing, the analysis of large volumes of data using artificial intelligence, and the Internet of Things (IoT), among other innovations [1]. These technologies are highly data-demanding and demand very high cybersecurity standards.

This paper aims to analyze the variables that explain the incorporation of cybersecurity procedures in companies in the Chilean manufacturing sector. To this end, several LOGIT and PROBIT models were built for binary dependent variables. Then, they were ordered with the information available in the Fifth Longitudinal Business Survey -ELE 5- of the National Institute of Statistics. This Survey includes a wide range of topics, which are grouped into i) Accounting and

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Finance, ii) Markets, Customers, and Suppliers, iii) General Management, iv) Resources, and v) Information Technologies. Our study exhaustively reviews the survey and selects 17 variables to analyze the factors explaining cybersecurity procedures' incorporation.

In this sense, the ELE 5 Survey identifies five types of internal cybersecurity procedures:

- Secure Password Authentication
- Identification and authentication of users through tokens or electronic devices (cards, USB, among others).
- Identification and authentication of users through biometric methods (fingerprint).
- Data backup (external hard drive, cloud computing) and
- Intrusion detection system (includes spam).

When reviewing the database of journals in the Web of Science Core Collection, Open Access, using the words "cybersecurity," "Adoption," and "Manufacture" to June 2023, we found that there are only sixteen publications, of which four are related to the analysis of specific technologies (IoT, 3D printing, Blockchain, BWM-CRITIC-TOPSIS). The rest is associated with the advancement of digital transformation and how cybersecurity plays a vital role within this process. It is noteworthy that this group of articles particularly emphasizes the importance of professional technical secondary education and the adoption of cybersecurity standards within the manufacturing sector [2]. Furthermore, when consulting the terms "cybersecurity," "Manufacture," and "Chile," we find only one study related to professional technical education. In conclusion, there are still few scientific publications that address, in a transversal and empirical way, with many companies, the levels of adoption of cybersecurity procedures.

The contribution of our work can be summarized in three axes:

- This research is innovative because it takes a data-driven approach to examine the adoption of cybersecurity practices in 574 Chilean companies. Our study encompasses various organizations and utilizes quantitative tools, specifically classical multivariate analysis techniques (Logit and Probit).
- Our work uses a technology adoption model to analyze the incorporation of cybersecurity practices in the company. We work on a database that addresses several company dimensions (marketing, organization, and management, among others). This allows our study to cross many different variables. Our research uses a validated and widely used model to analyze technological adoption (TOE).
- Finally, we examine the factors explaining cybersecurity adoption in a Latin American country, where digital transformation gaps are more pronounced compared to first-world countries [3]. We believe that the manufacturing reality in Chile is reflective of other developing countries.

We currently need to get information on this level of depth after COVID-19. However, manufacturing companies did not change their production practices for the most part, adjusting them according to social distancing health measures; therefore, our analysis includes structural elements that may be useful for a public policy to promote digital transformation.

II. THEORETICAL FRAMEWORK

Two main axes of analysis are developed. First, we present the concept of cybersecurity and its implications in Industry 4.0. Then, the generic technology adoption model is reviewed in a second point.

A. CYBERSECURITY AND INDUSTRY 4.0

Cybersecurity is one of the main challenges for those companies that want to become in Industry 4.0 [4].

According to Spadafora [5], manufacturing companies are the second most attacked in the USA due to the lack of security controls that protect their information assets according to the model: Confidentiality, Integrity, and Availability (CIA). Attacks through CAD files or USB disks with a malicious program or malware can affect the confidentiality and availability of data, and hackers who seek to compromise the integrity of the control interfaces employed by a human operator stand out.

The situation in Chile is the same. According to the study conducted by IPSOS in 2019 [6], 4 of 10 Chilean companies admitted to having had a cyber-attack. Furthermore, 40% of micro, 45% of small, and 56% of medium enterprises are considered vulnerable. In the cybersecurity guide of the Government of Chile of 2021, it is pointed out that Chilean companies have mainly suffered attacks of phishing, smishing, and ransomware, which have caused the temporary or permanent loss of files and access to services, the deletion of their websites or the Disruption of programs or systems.

As a result, companies have suffered from business continuity problems. Several studies highlight that the need for more awareness, training, and implementation of an adequate management system for information security are the main problems for small and medium-sized enterprises (SMEs) to be preferred to be attacked. In general, SMEs believe that investment in cybersecurity is an expense and do not see it as an ally that can mitigate business continuity problems and enhance the objectives and reputation of the organization.

In this sense, several regulations have been established to stimulate the adoption of cybersecurity in the industry, highlighting ISO 27.001 [7] and the NIST framework for cybersecurity [8]. These show the importance of maintaining adequate risk management on the company's information to continuously evaluate vulnerabilities, threats, and impact on information assets.

Another example of a standard even less known in Chile but no less relevant is the ANSI/ISA-62443, which corresponds to a series of standards for the safety of industrial control and automation systems. This standard emphasizes the application of the principle of defense in depth in the areas of plant safety, network, and system integrity. It is taking great value in the electricity sector.

In general, all the regulations, as mentioned earlier, promote the application of the following procedures [9]:

- Define the inherent vulnerabilities of systems that affect their security;
- Definition of cyber threats to systems.
- Identify risks related to cyberattacks.
- Countermeasures to address cybersecurity issues and subsequent assessment of residual risk obtained after implementation.

These steps require establishing an inventory of information assets to which a risk assessment is applied, considering confidentiality, integrity, and availability. This exercise must be updated regularly as a continuous improvement procedure, which allows an organization to have a vision of the current state of its level of maturity in cybersecurity, establish policies, procedures, improvement activities, and subsequent evaluation.

Considering the specific challenges and observations in small and medium-sized manufacturing companies regarding

cybersecurity, the facets of cybersecurity in the data-driven look can be further discussed as follows:

- a) Supply chain security: It is crucial to highlight the potential consequences of data breaches or unauthorized access to shared data. Many attacks are originated from associated suppliers, vendors, or third-party partners.
- b) Industrial Internet of Things (IIoT) security: The proliferation of connected devices and sensors in manufacturing environments brings new cybersecurity challenges and increases the landscape attack.
- c) Employee training and awareness: Small and mediumsized manufacturing companies often have limited resources for cybersecurity personnel. Therefore, it becomes crucial to invest in comprehensive cybersecurity training and awareness programs for employees, even in companies with limited resources. Nowadays, phishing is one of the most popular attacks due to the lack of cyber awareness.
- d) Incident response and business continuity: In the event of a cybersecurity incident, small and medium-sized manufacturing companies need to have effective incident response plans in place. This plan will help them to make good decisions as fast as possible and will help them to recover later quicker.
- e) Compliance with industry standards: Small and medium-sized manufacturing companies may be subject to industry-specific cybersecurity standards or regulations. Complying with relevant standards, such as the NIST Cybersecurity Framework or ISO 27001, can become mandatory regulations in some industrial sectors. This involves conducting risk assessments, implementing appropriate controls, and regularly reviewing and updating security practices to align with industry requirements.

B. FACTORS THAT MAY DETERMINE THE ADOPTION OF CYBERSECURITY PRACTICES

A review of the WOS-Core Collection database for "cybersecurity," "Adoption," and "Manufacture" to June 2023 yielded 16 results. A summary of this group is the ten most relevant articles shown in Table 1. In summary, the following emphases emerge:

- In relation to the Context: A first group of articles deals tangentially with cybersecurity, framing it within a broader theme associated with the implementation of enabling technologies for Industry 4.0. [10], [11], [12]. A second group accounts for cybersecurity and the development of some specific technologies, e. g. IoT, 3D printing, Blockchain, Digital Twin, among others. [13], [14], [15], [16], [17], [18]
- Regarding the results: A first group proposed different methodologies or tools to detect anomalies and attacks [13], [14], [15], [18]. Meanwhile, we have a second group that identified barriers or risks in implementing Industry 4.0 enabling technologies, among which is cybersecurity. [10], [11], [12], [16], [17]. In this regard

one research identifies barriers to adopting cybersecurity [19], using a survey of 258 organizations.

3) Regarding the method: In general, there is a group of articles that work with the expert consultation methodology using interpretive structural modeling (ISM), Analytical Hierarchy Process (AHP), among other methodologies. [11], [12], [16]. On the other hand, there are works oriented to explore a set of techniques or tools through simulation or cases [13], [14], [14], [14], [14], [18]. In this sense, the works that use sustainability reports [10], surveys of organizations [19], and literature review plus cases to generate a diagnosis [17] stand out.

Our research analyzes the explanatory factors of adopting some cybersecurity practices in 574 Chilean manufacturing companies, using multivariate techniques, complements ongoing research on the subject.

Many models allow us to understand the processes of technological adoption at the level of companies and end users. Our work will use the Technology, Organization, and Environment (TOE) model, which is more focused on the Diffusion of Innovation [20] because it is commonly used to understand the adoption process in companies.

Own elaboration from the Bibliographic review

The TOE model we will use is generic and, in future research, should be adjusted to cybersecurity because it is a business decision where leadership, policies, and the perception of cyber risk are fundamental [21]. In addition, we need to include cultural factors because they are vital when analyzing the adoption level of cybersecurity practices [22]. However, despite the above, in this research, we will stay with the homogeneous TOE given the quantitative nature of our study, where 574 companies are analyzed from 17 variables, generating quantitative indicators from the information available in the Survey.

Quickly, the dimensions of the TOE model are:

- Technological: Relative advantages provided by technology due to the perception of challenges and compatibility problems.
- Organizational: Firm size, management support, and workforce qualification.
- Environmental: The level of rivalry that the adopter company has, the levels of environmental uncertainty, and the perception of logistics support.

In a quick literature review, the ability of manufacturing companies to adopt cybersecurity practices can be explained by the following generic factors:

- The size of the company determines the capacity to adopt more complex technologies through the financial and administrative power to be able to mobilize scarce resources [19], [23], [24], [25], [26].
- Skilled labor facilitates the process of adoption of new technologies. This determines the company's prospecting, evaluation, and implementation processes [28], [29]. This is especially true in the workforce specialized in digitalization [21], [30], [31], which must have a

Article	Context	Proposal / key results	Method
[13]	Cyber-	Statistical methods of	Simulation and use of
	attacks and IoT use	attack detection.	sorting algorithms.
[10]	Industry 4.0	The level of presence	Examines 1501
[10]	and	of 4.0 technologies is	sustainability reports
	sustainability	broken down by type,	(GRI) to see the level
	reporting.	region, and impact on	of adoption.
	1 0	processes.	1
[14]	Increased	The neural network	Deep Learning is
	interconnecti	model allows for	proposed for threat
	on of devices	different types of	detection in IoT
	increases	anomaly detection	systems.
	cyber security	models.	
	requirements.		
[11]	Risks for	Risks are identified:	Survey of industry
[]	SMEs when	financial, operational,	experts in India using
	incorporating	technological, business,	Analytical Hierarchy
	Industry 4.0	social, supply chain,	Process.
	technologies.	and cybersecurity (data	
		breaches, hacking,	
101	Thereit	repudiation of attacks).	259
[19]	Identification of barriers to	Seven barriers were identified. They include	258 organizations from various sectors in
	adoption	lack of expertise, little	Australia and
	around	recognition in	international
	cybersecurity	implementation, little	organizations were
		government support,	surveyed. Response
		and that it is not a	distribution analysis
		driver for purchasing	and semantic analysis.
[10]	D estant	decisions	The second second second
[12]	Factors	Perception of benefits	Expert consultation,
	conditioning the	and management support is critical.	literature review, and interpretive structural
	implementati	Stresses the importance	modelling (ISM).
	on of	of having a 360°	
	Information	approach to cyber	
	and Digital	security.	
	Technologies		
	(IDT) in the		
	smart factory.		
	lactory.		
[15]	Expansion of	Implement various	Simulation and use of
	additive	Machine Learning	various algorithms.
	manufacturin	(ML) tools in additive	
	g (AM)	manufacturing	
		techniques.	
[16]	Use of	The blockchain is an	Literature review and
	Blockchain technology in	excellent tool to improve trust in a	interview with four project managers to
	global supply	global supply chain.	discover the state of
	chains	Essential knowledge to	Blockchain adoption in
		generate a decision	three third-party
		roadmap.	logistics services.
[17]	Application	Guidelines and	Literature review and
	of	recommendations for	case study, SWOT
	Blockchain	the implementation of	diagnosis.
	technology to	the Blockchain in the	
	the	automotive industry.	
	automotive		
	industry.	Visualize and	Development of
101			
[18]	New		multiple componente
[18]	New technologies	implement an	multiple components based on Artificial
[18]	New technologies (ML, IoT)	implement an integrated platform for	based on Artificial
[18]	New technologies	implement an	
[18]	New technologies (ML, IoT) impose	implement an integrated platform for simulating and	based on Artificial Intelligence. Textile

TABLE 1. Synth	hesis of the	e main sel	ected articles.
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cybersecurity culture [22], especially in the SME segment [19].

- Companies with an innovative track record are more likely to adopt new technologies for leadership, innovation culture, and organizational flexibility [25], [32], [33], [34].
- Companies with more compatible and interconnected technologies will likely adopt synergistic technologies. In digitalization ecosystems, cybersecurity is a strategic part of Industry s 4.0 [35]. The units that have a greater incorporation of 4.0 technologies and, at the same time,

have a more significant presence of these in the different stages of the value chain will have a greater probability of taking advantage of new opportunities in terms of speed, production capacity, reduction of errors, costs and an improvement in the quality and differentiation of the products [36]

- The existence of young, knowledge-intensive companies based on new technological fields increases the likelihood of adopting new technologies [37]. In this sense, the type of entrepreneur is fundamental to leading the digital transformation process, which was confirmed by Maggi C. [31] and Motta [30] in manufacturing SMEs in Chile and Argentina.
- Finally, access to some markets can pressure the adoption processes of specific technologies. Openness to international trade, whether in sales or purchases, is a stimulator of adopting 4.0 technologies [38]. Even trading with other countries can force companies to have a certain cybersecurity standard [21]. Certification in compliance with cybersecurity criteria could mean an advantage when attracting new customers or maintaining current ones. This issue depends on the sensitivity of consumers to the cyber risks associated with the misuse of their information [19].

In summary, the adoption of cybersecurity practices will be positively associated with the size of the company, the qualified human capital, the innovative trajectories of the organization, the existence of compatible technologies, youth, the origin of business capital, and the presence of markets that stimulate the existence of certain key technologies.

In this sense, the CSIRT of the Chilean government published a report on manufacturing companies [39]. This study presents global data on the main cyber security threats affecting this productive sector. Some examples of affected companies are given, but Chilean companies are not included.

The Global Cybersecurity Index (ITU) [40], which considers legal, technical, organizational, development, and cooperation aspects, ranks Chile 74th worldwide. Specifically, our country ranks seventh in the Americas, following the USA, Canada, Brazil, Mexico, Uruguay, and the Dominican Republic. Chile boasts an acceptable legal framework but faces significant weaknesses in technological aspects.

Since March 2018, Chile has had a Computer Security Incident Response Team (CSIRT) operating under the Ministry of Interior and Public Security. This team is responsible for strengthening and promoting good practices, policies, laws, regulations, protocols, and cybersecurity standards throughout the State, critical infrastructure, and the entire country (https://www.csirt.gob.cl/). Additionally, the draft Framework Law on Cybersecurity and Critical Information Infrastructure is currently in the Chamber of Deputies, undergoing its second constitutional procedure. This legal framework aims to enhance the institutional framework by establishing the National Cybersecurity Agency in Chile. Lastly, Law 19.628 on the Protection of Privacy, which dates back to 1999, is currently undergoing updates.

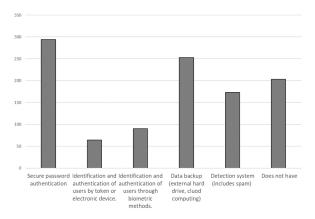


FIGURE 1. Frequency of occurrence of cybersecurity procedure. Number of enterprises = 574 (Note = a company can have multiple procedures). Own elaboration from the ELE.

Chile must urgently close the cybersecurity gaps to approach the standards of the most developed countries.

In this context, our research aims to analyze the factors influencing the adoption of cybersecurity procedures in the Chilean manufacturing sector, which contributes 9.7% to the Gross Domestic Product (GDP), ranking below mining activity (14.7%) and personal services (12.1%), according to data from the Central Bank of Chile. Within the manufacturing sector, notable industries include the food industry (31% share in manufacturing), the chemical industry (24%), and metal production (16%).

Chilean manufacturing plays a crucial role in the national export chains, serving as a strategic supplier to mining, forestry, and fishing activities. Consequently, cyber threats targeting individual manufacturing companies' operating networks can affect the entire production network, impacting both forward links (with customers) and backward links (with suppliers).

In this regard, we note an explosive increase in vulnerabilities in industrial control system (ICS) platforms [39], which increased by 49% when comparing 2020/2019. More specifically, in operational technology environments, during 2020, we have ransomware attacks (extortion software), which account for 33% of cases, remote access Trojans (RATs), which account for 15%, and internal incidents (associated with malicious insiders and negligence) account for 13% of the total.

Our work aimed to study cybersecurity as a critical link for national economic growth. With our study we sought to provide information necessary for new public policies, in the context of a new institutional framework f for promoting and developing cybersecurity at the national level.

Next, a pre-data review will be carried out to contextualize the quantitative analysis.

III. PRELIMINARY REVIEW OF DATA

About 35% (n = 203) of the companies analyzed need built-in cybersecurity procedures. This percentage is high considering the type of companies being analyzed, where manufacturing companies should have a minimum cybersecurity

standard due to the more significant addition of value in their production processes.

The most frequent procedures are strong password authentication, present in 51% of companies (n = 294); data backup on external disks or the cloud is present in 44% of companies (n = 253); and intrusion detection systems (including SPAM) were found in 30% of companies (n = 173).

In the range of the least frequent procedures, we have the identification and authentication of users through biometric methods, present in 16% of manufacturing companies (n = 90), and the identification and authentication of users by a token or an electronic device available in 11% of companies (n = 64).

A review of Table 1 shows that the Survey prepared by the National Institute of Statistics (INE) distinguishes different sizes of companies based on sales revenue. According to our tabulation, 33% of the companies are classified as significant (n = 187), and the medium ones represent 6% of the total analyzed (n = 37). Furthermore, in the group of small companies, the INE separates this segment into two sections, explaining 42% of the total (n = 243). Finally, the microenterprise segment represents 19% of the analyzed group (n = 107).

In another area, we see that, on average, companies are 24 years old. 40% of the units analyzed (n = 227) are family businesses. 29% of the organizations analyzed (n = 164) have invested in computer equipment in the last two years. Only 11% of all companies (n = 66) have invested in software in the last two years. The primary customer is the domestic market accounting for 60% of sales. While the international market explains an average of 7%, and sales to the State or public sector explain 3.9%. In the last two data, the dispersions are high from the coefficient of variation (301% and 345%, respectively).

Regarding its innovative dimension, 25% of the companies analyzed (n = 144) declare to participate in a trade association with a university and the productive development system. In this line, 26% of companies (n = 151) declare to participate in an R + D + i project. On average, companies hire 1,713 workers annually with a high dispersion (coefficient of variation of 474%); in monthly terms, the average number of workers is 142. However, only 29.7% are qualified as specialized workers, understood as managers, professionals, technicians, and qualified operators.

Finally, we have the technological variable. On average, companies use software other than traditional office software. In addition, 78%, equivalent to 449 units, use the Internet to interact with customers. However, 34% of these manufacturing companies (n = 194) use social networks, and only 36% of the organizations (n = 208) perform e-commerce operations (purchase or sales).

IV. METHODOLOGY

Our database comes from the Longitudinal Survey of Companies [44], which is carried out by the National Statistics Institute (INE) of the Ministry of Economy of the Government of Chile. This instrument aims to characterize the

Variable	Variable	Sum	Average	Deviation	C. variation
	type	Sum	Average	Deviation	variation
Company size					
Large company	Binary	187			
Median	Binary	37			
Small 1	Binary	77			
Small 2	Binary	166			
Microenterprise	Binary	107			
Antiquity	Numerical		24,7	14,6	59%
Family businesses	Binary	227			
Invested in computer equipment in the last two years Invested in	Binary	164			
software in the last two years	Binary	66			
% national sale	Numerical		59,9	42,6	71%
% foreign sales	Numerical		7,0	21,0	301%
% sale to the State	Numerical		3,9	13,3	345%
Number of major suppliers	Numerical		29,1	162,1	557%
Participation in Promotion System	Binary	144		EE	
Amount of complex software (excluding office)	Numerical		0,9	1,2	128%
Does R+D+i	Binary	151			
The annual number of workers	Numerical		1713	8121	474%
% Skilled workers	Numerical		29,7	31,9	107%
Use the Internet to interact with customers The company uses RRSS.	Binary Binary	449 194			
The company buys and sells using the Internet	Binary	208	ATION FROM		

TABLE 2. Presentation of the variables.

NOTE: OWN ELABORATION FROM THE ELE

country's companies and uses the database of the Internal Revenue Service (SII) as an input to determine the target population. To classify economic activities, it uses the International Standard Industrial Classification (ISIC4.cl.2012). The survey is answered by the owner or manager of the company and is a validated instrument, which has been applied since 2009, being a very relevant input for public policies in Chile.

Our model uses 17 explanatory variables from different company domains (Xn). The dependent variable (Yn), related to the adoption of cybersecurity procedures, arises from the same ELE 5 survey and is related to question J.III, on ICT security, and reads: "Does your company use one of the following internal security facilities or procedures? The respondent could tick more than one of the following alternatives:

- a) "Secure password authentication".
- b) "User identification and authentication via token or electronic device (e.g. USB card)."

TABLE 3. Distribution of the dependent variable.

	Number of	
Dependent variable	companies	%
Has cybersecurity procedure	371	65%
No cybersecurity procedure	203	35%
Overall total	574	100%
	Number of	
	companies	%
Null incorporation (0 procedure)	203	35%
Basic incorporation (1		
procedure)	124	22%
High incorporation (2,3,4,5		
procedures)	247	43%
Overall total	574	100%

Note: own elaboration from the ELE

- c) "Identification and authentication of users through biometric methods (fingerprint)".
- d) "Data backup (external hard disk, cloud computing).
- e) "Intrusion detection systems (including spam)".
- f) "Does not have "

The analysis assumes this database as a limitation of the field study. Since the survey was taken by the National Institute of Statistics, it is forbidden to reveal the name of the companies that provided the information (Organic Law 17.374). This limits the depth of the data of the surveyed enterprises within a specific subgroup. In the future, the quality of the measurement of cybersecurity in the company can be improved, which would be the subject of further research. Scales were generated from the responses.

We chose to binary the dependent variable based on the presence of cybersecurity procedures (0/1), considering that 35% of the companies do not have any procedures.

To better measure the depth of the adoption process, three scales of adoption were identified. From the results, a distinction was made between a) No adoption (0 procedure), b) Basic adoption (1 procedure) and c) High adoption (2 to 5 procedures). These scales are consistent with the cut-off points detected in the ordered Logit models and especially in the ordered Probit (see Table 5).

Measuring adoption from the sum of the procedures present in a company can be a basic proxy for technological depth. However, it is a first step to identify interrelationships between different technologies present in an organization [41]. This will be the subject of future research.

Four models were generated to identify the variables that best explain the level of adoption of cybersecurity procedures:

- Two binary models (Logit and Probit) to identify the factors that explain the presence or absence of cybersecurity procedures.

- Two ordered models (Logit and Probit) to identify the factors that explain the depth of the adoption process.

Regression models have the following formulation:

TO		¥7	
E	Variable	Var. type	Explanation and hypothetical relationship
Tech	Invested in		The company invested in assets during the
nolo gy	computer equipment.	Bina	previous year. Si=1; No =0 Positive hypothetical ratio (+)
53	Last two years	ry	[32], [33], [25], [34]
			The company invested in assets during the
	Invested in	D.	previous year. Si=1; No =0
	software. Last two years	Bina ry	Positive hypothetical ratio (+) [32], [33], [25], [34]
	two years	19	Total number of software (ERP; Sales,
	Amount of		marketing, and customer management software;
	complex		Turn-specific software, cloud computing,
Tech	software (excluding	Num e-	others) Positive hypothetical ratio (+)
nolo	office)	rical	[35], [36]
gy	Use the		The company declares to use the Internet to
	Internet to		interact with its customers. Si=1; No =0
	interact with	Bina	Positive hypothetical ratio (+)
	customers	ry	[35] [36] The company declares to use social networks.
			Si=1; No =0
	The company	Bina	Positive hypothetical ratio (+)
	uses RRSS.	ry	[35] [36]
	The company		During the year, the company performed buying and selling operations online (at least one).
	buys and sells		Si=1; No =0
	using the	Bina	Positive hypothetical ratio (+)
	Internet	ry	[35] [36]
			The total number of people employed during the
	The annual	Num	year (Jan+Feb++Dec). Direct contract (written or fee) included
	number of	e-	Positive hypothetical ratio (+)
	workers	rical	[23], [27], [24], [25], [26]
		Cata	The survey distinguishes the Large, Medium,
		Cate go-	Small 1, and Small 2 companies. Positive hypothetical ratio (+).
Orga	Company size	rical.	[23], [27], [24], [25], [26] [19]
nizat			=(Sum (managers + professionals + technicians)
ion	07 01-111-1	Num	/ total workers)*100%
	% Skilled workers	e- rical	Positive hypothetical relationship. [29], [30], [31], [21], [28], [19]
	workers	mean	Years of the company's existence from
			initiating activities in the Internal Revenue
		Num	Service.
	Antiquity	e- rical	Negative hypothetical ratio (-) [37]
	1 million of the second s	Tieur	A family or household owns more than 51%.
			Si=1; No =0
	Family	Bina	Negative hypothetical ratio (-)
	businesses	ry	[37] The percentage declared by the company of its
		Num	income sold to domestic companies.
	% national	e-	Positive hypothetical ratio (+)
	sale	rical	[38], [21], [19]
		Num	The percentage declared by the company of its revenue sold to international companies
	% foreign	Num e-	revenue sold to international companies. Positive hypothetical ratio (+)
	% foreign sales		revenue sold to international companies. Positive hypothetical ratio (+) [38], [21], [19]
Envi		e- rical	revenue sold to international companies. Positive hypothetical ratio (+) [38], [21], [19] Percentage declared by the company of its
Envi ron	sales	e- rical Num	revenue sold to international companies. Positive hypothetical ratio (+) [38], [21], [19] Percentage declared by the company of its income that is sold to the public sector.
Envi ron ment		e- rical	revenue sold to international companies. Positive hypothetical ratio (+) [38], [21], [19] Percentage declared by the company of its
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 TABLE 4. Presentation of variable and hypothetical relationships.

Adoption level = f (size; seniority; family businesses; investment in computer equipment; investment in software; % of national sales; % of foreign sales; % sale to the State; the number of significant suppliers; participation in trade associations, universities, and development system; the amount of complex software; does R + D + i; the annual number of workers, % skilled workers; use of the Internet to interact with customers; use of Social Networks; carrying out buying and selling transactions, (θ)

Appendix A presents the nonparametric correlation matrix where Spearman's Rho coefficient and Kendall's Tau correlation coefficient are calculated from the nature of the data [42]. It can be verified that no very high correlation generates suspicions of multicollinearity. On the other hand, the Variance Inflation Factor (VIF) is calculated, yielding results below 10.0, ruling out collinearity problems.

The original database of the Longitudinal Survey of Companies (ELE) registered 656 companies in the filtering process; companies that had incomplete and erratic data were excluded, reaching 574 units, equivalent to 87.5% of the original base.

Regarding the size restrictions to run the Logit and Probit models, Freeman's formula was used: [n = 10 * (k + 1)] [43], where k are the independent variables used (k = 17), then $n = 10^*(17+1)$; n = 180 companies required, our database reached 574, exceeding what was necessary.

Along with the ordered Logit and Probit models, marginal effects were also calculated to clearly identify each independent variable's impact when changing each dependent variable's tranche. Finally, the econometric analyses were done with the free software Gretl [45]. For the estimation of marginal effects, an lp-mfx package add-on, version 1.0, was downloaded.

V. FIELD STUDY

Now, we will present the results ordered by TOE dimension from Table 3. For each dimension and variable, we will offer the following:

- 1. Binary Coefficient Logit (Blb).
- 2. Binary Probit coefficient (Bpb).
- 3. Ordered Logit coefficient (Blo).
- 4. Ordered Probit coefficient (Bpo).

For each variable, the confidence level is presented with an asterisk. The rates of correctly predicted cases exceed 68%, indicating excellent models' good explanatory capacity (Table 4).

A. TECHNOLOGY

In all the regressions, companies that invested in computer equipment during the previous year have a greater probability of having cybersecurity procedures (Blb = 0.68^{**} , Bpb = 0.395^{**} ; Blo = 0.775^{***} ; Bpb = 0.441^{***}), confirming our initial hypothesis. However, the result was not as expected when we analyzed the investment in software. In all four regressions, this variable did not turn out to be significant, which rejects our preliminary hypothesis.

The technological variable "the number of complex software" is significant and positive when explaining adoption, which is tangible in all regressions (Blb = 0.46^{**} ; Bpb = 0.249^{**} ; Blo = 0.643^{***} ; Bpo = 0.365^{***}). The above confirms the technological compatibility with our initial working hypothesis.

In binary regression models, the Internet use to link with customers is not a significant variable (Blb = 0.28; Bpb = 0.18), which contradicts our working hypothesis. However, when we analyze the ordered models, using the Internet to link with customers is a positive and significant variable, confirming our initial assumptions (Blo = 0.421^* ; Bpo = 0.301^{**}). Furthermore, the variable use of the Internet to link with customers shows its importance when we identify different stages of adopting cybersecurity procedures.

The use of social networks in all models explains the adoption of cybersecurity procedures (Blb = 0.84^{***} ; Bpb = 0.475^{***} ; Blo = 0.901^{***} ; Bpo = 0.503^{**}). This relationship is positive and significant, confirming our working hypothesis.

Finally, we have the results of performing buying and selling operations online. Generally, this variable has a positive and significant relationship with adopting cybersecurity procedures (Blb = 0379^* ; Blo = 0.421^* ; Bpo = 0.179*). However, the results are not as strong as the previous variables. It is striking that in the Probit binary model, the purchase and sale over the Internet is not a significant variable to explain the adoption of cybersecurity practices.

B. ORGANIZATION

The first variable is the number of workers who behaved differently in our regression models. In the case of the binary Logit and Probit models, it is not observed that this variable is explanatory of adopting cybersecurity procedures. However, when we relate the variables in the ordered models, the number of workers becomes a vital indicator adjusting to our initial hypothesis (Blo = 0.00^{**} ; Bpo = 0.00^{***}).

When the size of companies is analyzed, we verify that large companies have a high probability of adopting cybersecurity procedures (Blb = $1,599^{***}$, Bpb = $1,007^{***}$; Blo = $1,351^{***}$; Bpo = 0.81^{***}). With a lower weight, mediumsized companies have an explanatory impact on adoption in all the regressions analyzed (Blb = 1.152^{**} , Bpb = 0.78^{***} ; Blo = 1.12^{**} ; Bpo = 0.69^{**}). When we examine the upper end of small businesses, there is still a probability of adoption of a lower significance (Blb = 0.609^{*} , Bpb = 0.37^{*} ; Blo = 0.54^{*}), and even for the ordered Probit model, the small business is no longer significant. Finally, smaller companies have no bearing on adoption. This gradient confirms the initial working hypothesis where the company size is unbalanced when explaining the adoption.

Contrary to our initial hypothesis, the variable percentage of specialized workers is insignificant in any of the four regressions. Something similar happens with the family condition of the company, where we do not see a significant impact on adoption in the four models built. Finally, although the age variable has a negative slope, it was non-significant.

C. ENVIRONMENT

Interestingly, in both models, the percentage of national sales presents a positive and significant result when explaining the adoption of cybersecurity procedures ($Blb = 0.007^{***}$, Bpb =

 0.004^{***} , Blo = 0.008^{***} , Bpo = 0.004^{***}), which confirms our hypothesis. Contrary to expectations, sales to foreign customers are not explanatory when adoption is worked as a binary variable. However, when we distinguish the depths of the adoption process, openness to international trade is a significant variable (Blo = 0.013^{**} , Bpo = 0.008^{**}), which allows us to confirm our initial hypothesis. Finally, when analyzing the dynamic capacity of sales to the public sector or the State, we find no significant impact on adopting cybersecurity procedures.

From the four regression models, we found that the variable "number of important suppliers" is insignificant in explaining adoption, which rejects our original hypothesis. Similarly, participation in trade associations, university consortia, and productive development systems is insignificant, rejecting our original hypothesis.

Finally, there is a significant and positive impact of those companies that carry out R + D + i and the adoption of cybersecurity practices (Blb = 0.892^{***} , Bpb = 0.495^{***} , Blo = 0.714^{***} , Bpo = 0.391^{***}), which proves our initial hypothesis.

Complementing the previous analysis, marginal effects are presented (Table 5) to identify any change in significance or slope in each depth section in adoption: a) Null incorporation, b) Basic incorporation, and c) Advanced incorporation. Next, the main changes for each dimension of the TOE model will be presented.

- **Technological dimension:** The investment in computer equipment in the last two years, the amount of complex software in the previous two years, customer service through the Internet, the purchase and sale by the Internet, and the use of the RRSS are positively and significantly associated when the level of adoption is advanced. However, consistent with previous results, investment in software in the last two years does not explain the adoption level.
- Organizational Dimension: The number of workers has a differential effect when the company is at an advanced incorporation level. Something similar happens with large and medium-sized companies where their condition is positively and significantly associated with high levels of incorporation of practices. Interestingly, in small companies, in the ordered Probit model, in none of the sections analyzed, any significant relationship is observed. Something different is the Logit-ordered model, where the incidence of small companies in the upper section shows that the advanced incorporation is very weak. In the case of smaller companies in this section, the relationship is nil. As previously concluded, the percentage of specialized workers, seniority, and family business status is not significantly related to the level of incorporation in any of its analyzed sections.
- **Environment Dimension:** The percentage of national sales, the percentage of sales abroad, and the realization of R+D have a positive and significant relationship

TABLE 5. Logit and probit regressions N = 574.

		(A) Binary logit (βlb)	(B) Binary probit (βpb)	(C) Logit Ordered (βlo)	(D) Probit Ordered (βpo)
TOE	Constant	-1,850 (***)	-1,064 (***)		
	Investment in computer equipment. Last two years (Binary)	0,683 (**)	0,395 (**)	0,775 (***)	0,441 (***)
	Software Investment. Last two years (Binary)	-0,203	-0,132	-0,313	-0,154
Г	Amount of complex software (office software excluded)	0,4618 (**)	0,249 (**)	0,643 (***)	0,365 (***)
	Uses the Internet to link with clients (binary)	0,2831	0,187	0,421 (*)	0,301 (**)
	The company uses the RRSS (binary)	0,8406 (***)	0,475 (***)	0,901 (***)	0,503 (***)
	Buy and sell online (binary)	0,379 (*)	0,198	0,345 (**)	0,179 (*)
	The annual number of workers	0,0004	-0,0002	0,0004 (**)	0,0002 (**)
	Large enterprise (binary)	1,599 (***)	1,007 (***)	1,351 (***)	0,811 (***)
	Medium Business (Binary)	1,152 (**)	0,785 (***)	1,120 (**)	0,691 (**)
	Small Business 1 (binary)	0,609 (*)	0,370 (*)	0,544 (*)	0,281
)	Small Business 2 (binary)	0,354	0,221	0,100	0,052
	Percentage of skilled workers.	0,00002	-0,0001	0,001	0,0004
	Antiquity	-0,006	-0,004	-0.007	-0,005
	Family business (Binary)	0,362	0,201	0,197	0,105
	% National sale	0,007 (***)	0,004 (***)	0,008 (***)	0,004 (***)
	% Foreign sales	-0,004	0,003	0,013 (**)	0,008 (**)
	% Sale to the state	-0,002	-0,001	0,002	0,00084
3	Number of major suppliers	0,001	0,0006	0,001	0,0007
	Participation in Trade Associations, Universities, and Productive Development Systems (Binary)	0,301	0,179	0,183	0,114
	Does R+D+i	0,892 (***)	0,495 (***)	0,714 (***)	0,391 (***)
	Cut-off points			C1=1,8(***) C2=3,5(***)	C1=1,0(***) C2=2,0 (***)
	Adjustment quality indicators	Number of correctly predicted cases = 79%	No. of correctly predicted cases = 78%	Núm.de correctly predicted cases = 69%	No. of correctly predicted cases = 68%
		R ₂ corrected=0.27 Chi ² .=0000	R ₂ corrected=0.27 Chi ² .=0000	Chi ² .=0000	Chi ² .=0000 Null hip (the erro
			Null hip (the error has normal distribution); Chi ² =1.30 p=0.520		has normal distribution); Chi ² =1.53 p=0.4

(*) 90% confidence; (**) 95% confidence and (***) 99% confidence.

Own elaboration from the ELE

in the section of companies with advanced incorporation. In both regression models ordered, the variables' percentage of the sale to the State, the number of suppliers, and participation in trade associations do not present a significant relationship in any of the sections analyzed.

From the slopes of marginal effects, we can conclude that the differentials in technological, organizational, and environmental variables are only clearly seen when companies have advanced incorporation. The analysis of marginal effects also shows us that in the face of changes in independent variables, the segment of null and basic incorporation companies has the same sensitivity.

VI. DISCUSSION OF RESULTS

From the results, we can conclude that:

Companies implementing a digital ecosystem [35],
 [36] will be more likely to adopt cybersecurity procedures. The above is visible in investment in computer equipment, complex software, relationships with

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customers through the Internet, electronic buying and selling, and the use of social networks. Moreover, this group of companies already knows the profitability of having a certain technological synergy, being more likely to implement or add new digital technologies by better visualizing the benefits, more excellent technological compatibility, and an innovative trajectory.

- A second strongly endorsed dimension is the size of companies as a determinant of adopting cybersecurity procedures. Our results align with those indicated by [19], [23], [24], [25], [26], and [27]. The number of workers hired, and the company's size from tranches derived from sales revenue show that large and medium-sized companies are more likely to mobilize financial, human, and organizational resources to implement new adoption processes.
- Contrary to expectations, the percentage of specialized workers turned out to be a variable that was not very significant, contradicting what was initially indicated by

TABLE 6. Marginal effects for ordered logit and ordered probit.

			Logit Ordered		Probit Ordered						
ТОЕ		Null incorporation of practices (dp/dx)	Basic incorporation of practices (dp/dx)	Advanced incorporation of practices (dp/dx)	Null incorporation of practices (dp/dx)	Basic incorporation of practices (dp/dx)	Advanced incorporation of practices (dp/dx)				
	Investment in computer equipment in the last two years (Binary)	-0.077 (***)	-0.105 (***)	0.182 (***)	-0.089 (***)	-0.080 (***)	0.170 (***)				
	Software Investment in the last two years (Binary)	0.038	0.039	-0.077	0.036	0.024	-0.06				
Т	Amount of complex software (office software excluded)	-0.07 (***)	-0.085 (***)	0.157 (***)	-0.081 (***)	-0.062 (***)	0,144 (***)				
	Uses the Internet to link with clients (binary)	-0.051	-0.052 (**)	0.104 (*)	-0.07 (**)	-0.04 (**)	0,119 (**)				
	The company uses the RRSS (binary)	-0.091 (***)	-0.120 (***)	0.212 (***)	-0.103 (***)	-0.090 (***)	0,194 (***)				
	Buy and sell online (binary)	-0.038 (**)	-0.046 (**)	0.084 (**)	-0.040 (*)	-0.030 (*)	0,070 (**)				
	The annual number of workers	-0.0005 (***)	-0.0006 (**)	0.00012 (**)	-0.0006 (***)	-0.0004	0,0001 (**)				
	Large enterprise (binary)	-0.129 (**)	-0.176 (***)	0.306 (***)	-0.156 (**)	-0.147 (***)	0,303 (***)				
	Medium Business (Binary)	-0.087 (**)	-0.151 (***)	0.238 (***)	-0.108 (***)	-0.137 (***)	0,246 (***)				
0	Small Business 1 (binary)	-0.052 (*)	-0.075 (*)	0.127 (**)	-0.056	-0.052	0,108				
	Small Business 2 (binary)	-0.011	-0.01	0.024	-0.011	-0.009	0,020				
	Percentage of skilled workers.	-0.0001	-0.002	0.0003	-0.0001	0.0007	0,0001				
	Antiquity	0.0008	0.001	-0.0019	0.0012	0.0009	-0.002				
	Family business (Binary)	-0.021	-0.026	0.048	-0.023	-0.018	0.041				
	% National sale	-0.0009 (***)	-0.001 (***)	0.002 (***)	-0.001 (***)	-0.00084 (***)	0,0019 (***)				
	% Foreign sales	-0.001 (**)	-0.001 (**)	0.003 (**)	-0.001 (**)	-0.0014 (**)	0,003 (**)				
	% Sale to the state	-0.0002	-0.0003	0.006	-0.0001	-0.0001	0,0003				
Е	Number of major suppliers	-0.0001	-0.00017	0.0003	-0.0001	-0.0001	0,0003				
	Participation in Trade Associations, Universities, and Productive Development Systems (Binary)	-0.019	-0.024	0.044	-0.024	-0.020	0,044				
	Does R+D+i	-0.070(***)	-0.097 (***)	0.168 (***)	-0.07 (***)	-0.072 (**)	0,151 (***)				

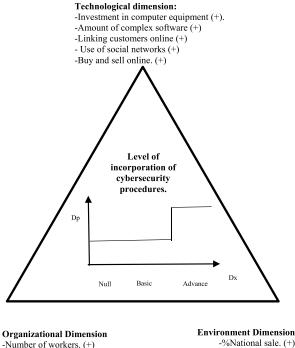
(*) 90% confidence; (**) 95% confidence and (***) 99% confidence Own elaboration from the ELE.

[21], [28], [29], [30], and [31]. Our study has not yet captured the importance of the workforce specialized in digitalization, especially those with cybersecurity training [19].

- The results reject the initial hypotheses concerning seniority and family business variables. One explanation for these results is that manufacturing companies that already have a level of complexity are being analyzed. Therefore, the type of firm does not contribute to the variety when explaining the dependent variable.
- The variable "percentages of national sales and international sales" are explanatory of the level of adoption, confirming the importance of the "pull of the markets" when explaining the adoption, which follows the line

of what was identified by [21] and [38]. However, it is worrying that sales to the State do not stimulate the adoption of cybersecurity procedures. This result shows a "gap" that must be covered by public policies where the State can promote processes of technical change in the companies that provide them as a strategy to boost the national productive fabric [19].

- In the environmental elements, it is striking that participation in Trade Associations, Universities, and Productive Development Systems does not explain the probability of adopting cybersecurity procedures. In this case, participation in these instances does not stimulate the implementation of cybersecurity procedures in manufacturing processes.



-Large companies (+)

-%National sale. (+) -% Foreign Sale. (+)

- Medium-sized companies (+)

-Company does R+d+i. (+)

FIGURE 2. Variables explaining the level of incorporation by dimension TOE. Source: own elaboration based on the results.

- A very different case is the companies that carry out R + D + i presenting high levels of incorporation of cybersecurity procedures. This significant and positive relationship shows the importance of innovative trajectories and having an organizational culture prone to technical change, which aligns with what is indicated in the theoretical framework.

The new enabling technologies of Industry 4.0 will continue to increase across different national productive sectors, owing to their numerous business benefits. The implementation of these new 4.0 technologies offers opportunities in six key areas [36]: a) enhanced flexibility through small batch production, b) accelerated prototyping, c) reduced set-up costs and fewer errors, d) minimized downtime, e) improved product quality and reduced rejected production, and f) better customer feedback on products. Howeve it is essential to note that these digital technologies, being highly data-intensive, are also more vulnerable to various cyber-attacks [39].

A significant part of the drivers of adoption of cyber security procedures identified in our study are replicated with other Industry 4.0 enabling technologies. Therefore, we will have a "co-evolution" of adoption trajectories between new digital production technologies and cyber security procedures.

In this context, manufacturing businesses must recognize the advantages of integrating digital technology into their production processes while also appreciating the benefits it brings to their company and the entire production network.

Securing data from the threats of cyber-attacks should be a top priority. Entrepreneurs must view cyber security not only as a necessity but also as a competitive advantage over their rivals.

The acceleration of co-evolution between industrial process enhancements and a higher standard of cyber security for company databases will be influenced by two key factors: firstly, the pressure from public policies in establishing standards, and secondly, the demands from larger companies acting as customers. These factors will play a significant role in adopting robust cyber security measures across the industry.

VII. CONCLUSION

The conclusions will be divided into three parts: regarding the methodology, regarding public policies, and regarding a new adoption model for cybersecurity.

A. THE METHODOLOGY AND LINES OF FUTURE RESEARCH

The first conclusion we can obtain is associated with the research limits. Agreeing with what was proposed by [21], the generic TOE model by itself does not allow us to understand the adoption of cybersecurity procedures because there are qualitative elements associated with the perception of those responsible for ICTs about the existing risks either by unknown, hypothetical and intangible threats and secondly, to the political dimension by establishing mandatory cybersecurity procedures, which must be respected by the different actors, which implies coordinating, educating, raising awareness or punishing if they are not fulfilled promptly.

It is pertinent that the surveys better record some critical elements in the cyber security of companies, for example: a) the existence of staff training in cyber security, b) the level of knowledge of ISO 27001 cyber security standards, c) the existence of procedures before, during and after cyberattacks, d) the presence of "risk maps" where the threat and impact of cyber-attacks on key assets is assessed, e) the frequency of backups, among other aspects.

Although the ELE 5 survey evaluates certain cyber security aspects such as the adoption of technologies to reinforce access control to the buildings of a company, in no case, did this survey asked if an organization carries out cybersecurity awareness and training for its workers. Nor does it address prevention aspects that would diagnose if workers can identify possible attempts at phishing, smishing, and ransomware attacks (the most reported attacks in Chile). Moreover, the survey does not determine whether manufacturing companies are interested in including cybersecurity standards and good practices beyond those regulated in their sector.

Upon reviewing the existing literature, we observed a scarcity of studies analyzing the adoption of cybersecurity among a group of companies. We believe that our research makes a significant contribution in this field for several reasons: i) Our study examines explanatory factors based on a comprehensive sample of 574 companies, utilizing classical

multivariate analysis methodologies (Logit and Probit), ii) It addresses the cybersecurity challenge by employing a well-established adoption model (TOE), providing valuable insights into the adoption process y iii) furthermore, our research delves into the cybersecurity concerns specific to a developing country, which faces technological gaps in comparison to more developed nations. This aspect adds a valuable dimension to understanding cybersecurity practices in diverse socio-economic contexts.

We identify the following lines of future research:

- It is necessary to look at the links between companies to identify the extent to which a large company (e.g., an exporter) puts pressure on an SME to adopt specific cybersecurity standards. Knowledge of the productive fabric will make it possible to estimate the diffusion speed of new technologies and cybersecurity procedures.
- It is relevant to know the interrelationships between technologies within companies. This implies identifying for each specific technology (e.g., SCADA, IoT; PLC, EWS, etc.) the level of risk of cyber-attack. Due to their high interoperability, it is interesting to identify how new 4.0 technologies can be a "gateway" for cyber attackers, affecting the entire operational network of the manufacturing company and its production environment.
- It is necessary to visualize where the SME can start to strengthen its cybersecurity (procedures and facilities). In our opinion, an interesting topic to address is the "technology trajectories" for each type of company, distinguishing the size, industrial sector, and strategic content of its production.

It is essential to establish international research networks focused on studying cybersecurity in the industry from a global perspective, aiming to: i) develop specific tools for identifying the current state of cybersecurity in manufacturing (e.g., surveys, incident reports, among others), ii) identify new common patterns in cyberattacks, and iii) determine new protocols for prevention, response, and business continuity.

Currently, various international organizations are involved in cybersecurity efforts, such as the United Nations (UN), the International Telecommunication Union (ITU), the Cybersecurity Center in Madrid (CCMAD), the National Institute of Cybersecurity in Spain (INCIBE), the Open Web Application Security Project (OWASP), the CSIRT of Chile, among others. There is a need to establish a global research network since cyber attackers do not respect traditional national boundaries.

B. REGARDING PUBLIC POLICIES AND ISO 27001

In the Chilean context, there is an emerging institutional framework for cybersecurity. Notably, the Computer Security Incident Response Team (CSIRT) was established in 2018, and currently, there are ongoing discussions in the Chamber of Deputies to enact the first Framework Law on Cybersecurity. Given this landscape, it becomes paramount to foster a "cybersecurity culture" [22] by promoting the

dissemination of procedures within organizations, integrating them into the fabric of production processes, and particularly among citizens. Citizens, in their roles as customers, can play a vital role in encouraging companies to enhance their security standards. Simultaneously, they can exert pressure on the State to improve regulatory frameworks that involve fines and sanctions. Instilling such a cybersecurity culture across different levels of society will create a more robust and secure digital environment, benefitting both the private and public sectors in Chile.

The State can also help disseminate and encourage the use of cybersecurity standards at the company level (e.g., tax incentives, public procurement systems, etc.) as a condition for accessing key customers in international markets, which is especially necessary for SMEs [19].

In this regard, the government, in collaboration with other key stakeholders including businesses and universities, should implement a National Cybersecurity Plan for manufacturing Small and Medium-sized Enterprises (SMEs). This initiative should commence with a comprehensive survey of the sector during its diagnostic phase and should culminate in public policies aimed at strengthening cybersecurity. This entails enhancing incident monitoring through the establishment of a new Manufacturing CSIRT (Computer Security Incident Response Team), fostering the development of new technology providers, providing training, implementing certification of standards (ISO 27001 and NIST), and raising awareness within each company.

Finally, ISO 27001 presents several guidelines for implementing an information security management system to manage risks and establish control measures to protect an organization's information assets, including the TOE model's dimensions.

- Technology. According to the standard, the use of technology is intended to support critical activities and that it will be able to allow the organization to meet its objectives. The standard promotes using cryptography for storing data in transit and at rest, using semi-automated technological equipment, processes, and physical security monitoring mechanisms. The use of good practices and the use and development of software are promoted, as well as establishing protocols that direct the continuity and recovery of an organization in the event of a security incident.
- Organization: ISO 27001 advises that the risk management process be wholly aligned with the culture, processes, structure, and strategy of the organization. Each member of the organization must have clear roles and accountability. There must be clear policies, standards, guidelines, and documented models known and practiced by its members. For this, senior management must demonstrate an exemplary and committed attitude, encouraging the continuous improvement of all the resources that support the organization.
- Environment: The standard highlights the importance that the organization knows the needs and expectations

of its stakeholders (employees, customers, suppliers, investors, etc.) and the social, cultural, legal, financial, and competitive environments at national and international levels where it operates. Must identify critical aspects of the business and trends that impact the organization's goals.

Although some aspects indicated by ISO 27001 were analyzed in this article, as future lines of work, we hope to deepen under its eaves in the study of the Chilean manufacturing industry according to the dimensions of the TOE model.

C. REGARDING AN APPROPRIATE ADOPTION MODEL

We believe the generic TOE (Technology, Organization, and Environment) model can adopt cybersecurity procedures when the following key factors are included:

- 1. **Technological factors** encompass the availability, maturity, and effectiveness of cybersecurity technologies and solutions. This includes the existence of robust security tools, frameworks, and standards that can be leveraged to protect an organization's technology infrastructure, systems, and data. The presence of advanced threat detection and prevention mechanisms, encryption technologies, and secure communication protocols can significantly influence the adoption of cybersecurity procedures.
- 2. **Organizational factors** affect an organization's internal structures, processes, and capabilities. These factors include dedicated cybersecurity teams, welldefined roles, responsibilities, and allocating sufficient resources and budget for cybersecurity initiatives. Additionally, the organization's commitment to creating a cybersecurity culture and fostering awareness among employees plays a vital role in determining the success of adopting cybersecurity procedures.
- 3. **Human factors** refer to the knowledge, skills, and attitudes of individuals within the organization. This includes the level of cybersecurity awareness, training, and education provided to employees. The willingness and motivation of employees to adhere to cybersecurity best practices, such as strong password management, regular updates, and vigilant behavior, greatly impact the successful adoption of cybersecurity procedures.
- 4. **Regulatory and compliance factors** encompass the legal and regulatory requirements that organizations must adhere to regarding cybersecurity. These may include industry-specific regulations, privacy laws, data protection requirements, and international standards such as ISO 27001. The need to comply with these regulations provides a strong incentive for organizations to adopt cybersecurity procedures
- 5. Environmental factors: consider the external influences on an organization's cybersecurity posture. This includes the evolving threat landscape, emerging cybersecurity risks, and the overall security climate within the organization's industry or geographical region. High-profile cyber incidents and public awareness

of cybersecurity issues can influence organizations to adopt cybersecurity procedures to protect against potential threats and reputational damage.

6. Economic factors: pertain to the financial aspects of cybersecurity adoption. This includes the cost of implementing cybersecurity measures, the return on investment (ROI) associated with cybersecurity investments, and the cost of potential breaches or incidents. Organizations must evaluate the financial viability and long-term benefits of adopting cybersecurity procedures about their overall business goals and risk tolerance.

Considering these factors within the TOE model, we believe organizations can assess their readiness and capacity to adopt cybersecurity procedures effectively. The interaction and alignment of technological, organizational, human, regulatory, compliance, environmental, and economic factors determine the level of cybersecurity maturity and the extent to which cybersecurity procedures can be adopted and integrated into an organization's operations.

APPENDIX A CORRELATION MATRIX FOR NONPARAMETRIC VARIABLES

	Kendall's tau correlation coefficient																					
						К	enc	'llat	s ta	u co	orre	lati	on d	:oef	ifici	ent						
	a)	b)	c)	d)	e)	f)	g)	h)	i)	j)	k)	1)	m)	n)	o)	p)	q)	r)	s)	t)	u)	v)
a)		0,0	<u> </u>	0,0		0,2	0,1	0,1	<u> </u>	0,2	0,1	0,3	0,0	0,0		0,0	0,3	0,0		-0,2	0,1	0,1
b)	0,0		0,0			-0,1	0,0		<u> </u>	· · ·		· · ·		- <u> </u>		0,0		· · ·	<u> </u>	-0,1	0,0	0,1
c)	0,1	0,0		0,4	,	0,2	0,2		<u> </u>	0,4		0,2	0,1	0,2	0,2		0,3	0,0		-0,2	0,4	0,3
d)	0,1	0,0			0,0	0,1	0,1	0,2	<u> </u>	0,3		0,2	0,1	0,2	0,1	0,1	0,3	0,0	-0,1	-0,1	0,2	0,2
e) f)	0,0	0,0 -0.1	0,1 0,2			-0,2	-0,1 0,1	0,0			0,1	0,1 0,4	0,1	0,1	-0,1 0.1	0,0 0,0	0,1 0,5	0,1 0,0	-0,1 -0,1	0,0 -0.2	0,2 0,3	0,2
g)	0,3	-0,1			-0,2	0,1	0,1	0,3	0,5	0,4	0,5	0,4	- <u>´</u>	0,1	0,1	0,0	0,5	0,0	0,0	0,0	0,5	0,5
<u>в/</u> h)	0,1	0,0	,	<u> </u>	<u> </u>	0,1	0,1		0,1	0,1		0,3	0,1	0,1	0,1	0,1	0,0	0,1	0,0		0,1	0,1
i)	0,2	0,0				0,3	0,1			0,4	0,2	0,3	0,1	0,2	0,1	0,2	0,4	0,0	-0,1	-0,2	0,3	0,3
J)	0.3	0,1	0,4	<u> </u>		0,4	0,1				0.3	0,6	0,2	0.2		0.1	0,6	0.1	-0,1	-0,4	· ·	0,4
k)	0,1	0,0				0,3	0,1	0,2				0.2	0,2	0,2	0,1	0,2	0,3	0,0	-0,1	-0,1	0,3	0,3
)	0,4	0,0	· ·			0,5	0,0	· ·	<u> </u>	· ·			0,1	0,2	0,1	0,1	0,6	0,1	0,0	· ·	0,5	0,4
m)	0,0	-0,1	0,2	0,1	0,2	0,2	0,1	0,1	0,1	0,2	0,2	0,1		0,2	0,0	0,1	0,1	0,0	0,0	0,0	0,2	0,1
n)	0,0	0,0	0,2	0,2	0,2	0,1	0,1	0,2	0,2	0,3	0,2	0,2	0,2		0,2	0,2	0,2	0,1	0,0	-0,2	0,3	0,3
o)	-0,1	0,0	0,2	0,1	-0,1	0,1	0,1	0,1	0,1	0,2	0,1	0,1	0,0	0,2		0,3	0,1	0,0	-0,1	0,0	0,3	0,2
p)	0,0	0,0	0,1	0,1	0,0	0,0	0,2	0,1	0,2	0,1	0,2	0,1	0,1	0,2	0,3		0,0	0,1	0,0	0,0	0,2	0,2
q)	0,4	0,0	0,3	0,3	0,1	0,5	0,0	0,4	0,4	0,7	0,3	0,7	0,1	0,2	0,1	0,0		-0,2	-0,3	-0,4	0,5	0,4
r)	0,0	0,0	0,0	0,0	0,1	0,0	0,1	0,1	0,0	0,1	0,0	0,2	0,0	0,1	0,0	0,1	-0,2		-0,1	-0,2	0,1	0,1
s)	0,0	0,1	-0,1			-0,1	0,0	0,0				0,0	0,0	0,0	-0,1	0,0	-0,3			-0,3		-0,1
t)	-0,2	-0,1	<u> </u>	-0,1		-0,3	0,0	<u> </u>	<u> </u>	_		-0,4	0,0		<u> </u>	- · ·	-0,4	-0,2	-0,3		-0,3	-0,2
u)	0,2	0,0				0,4	0,1	0,3	<u> </u>	· · ·			0,2	0,3	· ·	0,2	0,6		-0,1	-0,3		0,8
v)	0,1	0,1	0,3	0,2	0,2	0,3	0,1	<u> </u>	L ć	L (<u> </u>	· ·	· ·	L /	<u> </u>		0,4	0,1	-0,1	-0,2	0,9	
							pea	irm	ans	co	rrela	atio	n co	еπ	icie	nτ⊵						
(a	a) Sen	iority									(k) D	oes I	₹+D+	i								
(1	b) Fan	nily b	usine	ess (E	inar	()					(L) Total number of workers											
(0	c) Inve	estme	nt in	com	pute	r equ	ipm	ent (I	Binar	y)	M) Percentage of skilled workers.											
(0	d) Inv	estme	ent ir	n soft	ware	e (Bin	ary)				(n) Uses the Internet to link with clients											
(e) % C	ome	stic s	ale							(binary) (o) The company uses the RRSS (binary)											
	f) % Fe										(P) Buy and sell online (binary)											
(g) % S	ale to	the	state	9						(q) Large company (binary)											
(1	h) Nui	nber	of m	ajor	supp	liers					(r) Medium-sized enterprise (binary)											
ù	i) Pa Jniver ystem		, an	nd P		rade ctive			atior pme		(s) Sr	nall	Busir	iess 1	L (bir	iary)						
(1	, J) Νι oftwa	ımbei	r of	со	mple	x so	oftwa	are	(offi	ce	(t) Sr	nall I	Busin	ess 2	! (bir	ary)						
											(u) D	iscre	et se	curit	y pra	ctic	es					
											(v) B	inary	secu	urity	pract	ices						

Own elaboration from the ELE.

APPENDIX B

• CIA: Security model, Confidentiality, Integrity, and Availability

- IoT: Internet of Things
- SCADA: Supervisory Control and Data Acquisition
- BWM: Best-Worst Method
- CRITIC: CRiteria Importance Through Intercriteria Correlation
- TOPSIS: Technique for Order of Preference by Similarity to Ideal Solution
- IPSOS: Independent Polling System Of Society
- ISO: International Organization for Standardization
- NIST: National Institute of Standards and Technology
- ANSI/ISA: American National Standards Institute Instrument Society of America
- TOE: Technology, Organization, and Environment model
- PLC: Programmable Logic Controller
- EWS: Exchange Web Services
- CSIRT: Computer Security Incident Response Team
- ROI: Return on Investments

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