# Ready or Not, AI Comes – An Interview Study of Organizational AI Readiness Factors

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Appendix (available online via <a href="http://link.springer.com">http://link.springer.com</a>)

## **Online Appendix**

## Appendix I: Theoretical Foundation

Table 2: Enterature Overview of Our Theoretical Foundatio	Table 2: Lite	rature Overview	of Our Theor	etical Foundatio
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Literature stream	Core insights	Key references		
Innovation and	- Innovation adoption is defined as a process with different phases	Chwelos et al. 2001; Davis 1985;		
technology adoption	- Within the process, several factors determine the adoption decision	Hameed et al. 2012: Oliveira and		
	- The research investigated factors and antecedents to technology adoption on an organizational level (e.g., DOI and TOE) or acceptance on an individual level (e.g., TAM and TPB)	Martins 2011		
	- Theory needs to be tailored to the adoption context			
Organizational	- Readiness is a state, which is attained before a specific activity takes place	Armenakis et al. 1993; Helfrich et		
readiness for change	- Precursor condition for implementation of a change such as innovation, technology, etc.	al. 2011; Rafferty et al. 2013; Shahrasbi and Paré 2014; Weiner		
	- Systematic reviews identify that the organizational readiness construct is not yet well-defined	2009		
	- Discussed factors comprise change valence, change efficacy, and contextual factors			
(Digital) readiness in IS	- The literature discusses readiness for technology adoption in the context of systems implementation, knowledge management, e-readiness, and Green IT readiness	Abdinnour-Helm et al. 2003; Lokuge et al. 2018; Molla and Licker 2005; Nguyen et al. 2019;		
	- Organizational readiness affects the probability of successful technology adoption	Snyder-Halpern 2001		
	- Readiness is a capability which requires continuous improvement			
	<ul> <li>Typical variables have been employed: financial resources, IT sophistication, management support, trading partner readiness, elapsed time since adoption, organizational culture, communication of goals, individual attitude, commitment</li> </ul>			
	- Digital readiness is conceptualized as digital assets, digital capabilities, and digital commitment			
AI readiness	- The literature on AI adoption and AI readiness is limited to a few qualitative studies	Alsheibani et al. 2018; Pumplun et al. 2019		
	- Two main contributions (Alsheibani et al. 2019; Pumplun et al. 2019) gather and list factors within the TOE model which determine AI adoption			

#### Appendix II: Overview of Interview Experts

ID	Interviewee Position	Industry	Company Size	AI Adoption Stage	Duration (minutes)
E01*	Senior Data Scientist	Automotivo	(100.000	Continued Use	50
E02*	Director Digital Transformation	Automotive	<100.000	Continued Use	53
E03	Chief Business Officer	IT	<10	Continued Use	58
E04	Managing Partner	Venture Capital	<50	Awareness	60
E05	IT Architect Enterprise Operations Center & Workload Automation	IT	>100.000	Continued Use	61
E06	Professor for Innovation and Technology Management	Research	<10	Awareness	50
E07	Head of Asset Intelligence Center	Logistics	<50.000	Evaluation	55
Eo8	Digital Advisor	Software	>100.000	Continued Use	70
E09	Senior Digital Expert	Manufacturing	<25.000	Intention	57
E10	Head of Strategy and Innovation	IT	<50	Continued Use	66
E11	Managing Director	IT	<50	Continued Use	52
E12	Member of the Executive Board	Insurance	<5000	Evaluation	49
E13	Head of Operations	Healthcare	<50	Continued Use	53
E14	Head of Functional Controlling	Pharma & Agriculture	>100.000	Continued Use	56
E15	Director	Automotive	<50	Intention	62
E16	Vice President Core Business Apps	Manufacturing	<25.000	Consideration	56
E17	CIO	Automotive	>100.000	Continued Use	79
E18	Head of IT	Construction	<100.000	Intention	54
E19	Head of Digital Unit	Manufacturing	<5000	Consideration	46
E20	CIO / Vice President IT	Automotive	<100.000	Intention	58
E21	CEO	IT	<50	Continued Use	58
E22	CEO	Automotive	<50	Continued Use	49
E23	CEO	Consulting	<500	Awareness	53
E24	Head of Center of Excellence IT Automation	Manufacturing	<25.000	Consideration	58
E25	CEO	IT	<250	Continued Use	72

#### Table 3: Overview of Interview Experts

\* Note that E01 and E02 are respondents from the same company and were interviewed in the same interview meeting

#### Appendix III: Explicating and Validating the AI Readiness Factors

For a comprehensive AI readiness assessment, we developed indicators with insights from our expert interviews, the literature, and practitioner studies. To the best of our knowledge, the research does not yet provide such detailed indicators. To ensure a rigor approach, we referred to established procedures for scale development (Boateng et al. 2018). Specifically, we paid attention to ensure unambiguity and simplicity and to prevent exceptionally lengthy measures (DeVellis 2017). Thus, our first draft of indicators resulted in 76 indicators that explicated the 23 preliminary AI readiness factors.

Finally, we gathered a focus group of seven AI-related researchers to perform a card-sorting procedure of our categories, factors, and indicators in a joint workshop. Card-sorting assesses the construct validity and identifies ambiguous measures (Moore and Benbasat 1991). Thus, the results helped us to validate our initial categorization and further improve our factors and indicators. We asked the focus group members to assign the randomized indicators to the list of factors and to rate their confidence level in the assignment. One author moderated and discussed personal feedback of the focus group during and after the assignment. This provided us with additional insights into focus group members' understanding and potential areas for improvement.

Overall, the card-sorting scored a hit ratio of 72.74%. Despite mostly satisfying results, the card-sorting also revealed weaknesses in some AI readiness factors (see **Table 4**). Particularly, the average hit ratios of data-related factors (62.43%) were lower compared to non-data-related factors (78.43%). While this is partly dependent on the focus group constellation, we also took measures to account for the received feedback. For instance, we improved the wording of some factors to improve comprehensibility. Further, we restructured all data factors to depict data criteria (i.e. characteristics and their management) instead of a distinction between data characteristics and data management factors. We also improved the indicators' wording to avoid ambiguities. For instance, we replaced the word 'aware' in some indicators to avoid unintended links to the factor 'AI awareness'. In rare instances, we dropped indicators altogether because of particularly low hit ratios or missing AI-specifics. This procedure led to our final compilation of 18 AI readiness factors organized in five categories and operationalized by 58 indicators (see **Table 5**).

#### Table 4 Card-Sorting Hit Ratios

	0	Ton	Are	-AS			7.	$\mathbb{Z}$		<u> </u>	G	Q		2		2	>				<u>&gt;</u>		abie .			
Factor	AND TO S	lip ta	64 9.7	1 28	and All	\$ 13	Ing way	4 C.	Ds sta	3 (Q	20 31	No Va	ha iai	00 44	no Oa	13 - 80.	13	Icc Da	610 Da	20 ° 84	are and	13 Ja	00 402	3	Total Sorts	% Hits
Comp*	9	1	0	7	0	0	0	6	0	0	3	0	0	0	0	0	0	0	0	0	0	0	0	2	28	32.14%
CuAiR	0	23	2	0	0	0	0	1	0	0	1	0	0	0	0	0	0	0	0	0	0	1	0	0	28	82.14%
ToMaSu	0	0	21	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	21	100.00%
ProAiAl*	0	0	0	18	0	0	0	0	0	0	2	0	1	0	0	0	0	0	0	0	0	0	0	0	21	85.71%
FinBud	0	0	0	0	21	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	21	100.00%
AiExp*	1	0	0	0	1	15	0	2	0	0	0	2	0	0	0	0	0	0	0	0	0	0	0	0	21	71.43%
ItInf	0	0	0	0	0	0	21	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	21	100.00%
AiAwa*	0	0	0	0	0	1	0	25	2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	28	89.29%
Ups	0	0	0	0	1	0	0	0	23	0	0	4	0	0	0	0	0	0	0	0	0	0	0	0	28	82.14%
EtaCo*	1	0	0	1	0	0	0	0	0	17	0	0	0	0	0	0	0	0	0	0	0	0	1	1	21	80.95%
Inno	1	0	0	0	0	0	0	1	2	0	17	3	1	0	0	0	0	0	0	0	0	0	0	3	28	60.71%
CollWo	1	0	2	1	1	2	0	0	0	0	1	13	0	0	0	0	0	0	0	0	0	0	0	0	21	61.90%
ChaMa	1	0	1	0	0	1	0	0	0	0	0	0	17	0	0	0	0	0	0	0	0	0	0	1	21	80.95%
IterDe*	0	0	0	0	0	0	2	0	0	0	0	2	0	29	0	0	0	0	0	0	0	0	0	2	35	82.86%
DaAmo*	0	0	0	1	0	0	0	0	0	0	0	0	0	0	2	3	0	6	1	1	0	0	0	0	14	14.29%
DaTy*	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	8	0	0	0	0	6	0	0	0	14	57.14%
DaQua	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	2	16	0	0	2	1	0	0	0	21	76.19%
DaAcc	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	26	0	0	0	0	1	0	28	92.86%
DaFlo	0	0	0	0	0	0	2	0	0	0	0	0	0	0	0	0	2	0	17	0	0	0	0	0	21	80.95%
DaCo*	0	0	0	0	0	0	0	0	0	0	0	0	0	0	8	0	0	0	0	12	1	0	0	0	21	57.14%
DaPre*	1	0	0	2	0	0	0	0	0	0	0	0	0	0	0	0	10	0	0	2	8	4	0	1	28	28.57%
DaAna*	4	0	1	0	0	1	2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	12	0	1	21	57.14%
DaPo*	0	0	0	0	0	0	0	0	0	3	0	0	0	1	0	0	0	0	0	0	0	0	17	0	21	80.95%

Abbreviations

Comp	Compatibility	ItInf	IT infrastructure	ChaMa	Change management	DaFlo	Data flow
CuAiR	Customer AI readiness	AiAwa	AI awareness	IterDe	Iterative development	DaCo	Data collection
ToMaSu	Top management support	Ups	Upskilling	DaAmo	Data amount	DaPre	Data preparation
ProAiAl	Process AI alignment	EtaCo	Ethics and compliance	DaTy	Data type	DaAna	Data analytics
FinBud	Financial budget	Inno	Innovativeness	DaQua	Data quality	DaPo	Data processing
AiExp	AI experts	CollWo	Collaborative work	DaAcc	Data accessibility		

\*Factors with changes after card-sorting (renamed, dropped, or restructured with other factors)

### Appendix IV: Overview of AI Readiness Categories, Factors, and Indicators

Abbreviation	Illustrative AI readiness indicators	Key References					
<u>Strategic</u>	AI adoption needs to be aligned with the overall strategy of an organization. As such, the category s	trategic alignment is defined as the tight linkage of					
Alignment:	organizational priorities and processes enabling and supporting this adoption process.						
	AI-business potential	<ul> <li>E01, E03, E08, E10, E13, E18, E22</li> </ul>					
AiBuPo1	My organization has business problems and opportunities that are suited to be addressed by AI.	• Alsheibani et al. 2018; Groopman 2018; Hofmann et al.					
AiBuPo2	AI technologies offer potential benefits for the organization's business.	2020; PA Office of Administration 2018; Pringle and					
AiBuPo3	My organization has appropriate methods and procedures for AI use case discovery.	Zoller 2018; Pumplun et al. 2019					
	Customer AI readiness	• E03, E08, E11, E24					
CuAiR1	My organization's customers are likely to accept AI-integrated offerings.	Groopman 2018; Pumplun et al. 2019					
CuAiR2	My organization's customers have adequate expectations toward AI-integrated offerings.						
CuAiR3	My organization's customers have error tolerance toward AI-integrated offerings.						
	Top management support	• E08, E10, E13, E18					
ToMaSu1	Top management integrates AI adoption into the strategy of the organization.	Alsheibani et al. 2018: Catalyst Fund 2020: Groopman					
ToMaSu2	Top management is willing to actively pursue AI adoption through initiatives and projects.	2018; Intel 2018; Kruse et al. 2019; Microsoft 2020; PA					
ToMaSu3	Top management is willing to support bottom-up AI initiatives.	Office of Administration 2018; Pringle and Zoller 2018;					
0		Pumplun et al. 2019					
	AI-process fit	• E09, E11, E16, E24					
PrAiFit1	My organization documents and standardizes business processes and operations.	Groopman 2018: Microsoft 2020: Pumplun et al. 2019					
PrAiFit2	My organization is willing to implement new business processes and operations to enable AI-	r i i i i i i i i i i i i i i i i i i i					
	integrated workflows.						
PrAiFit3	My organization is willing to reengineer business processes and operations to enable AI-						
	integrated workflows.						
	Data-driven decision-making	• E10, E15					
DaDriDM1	My organization is capable to create insights with data-driven analytics.	Catalyst Fund 2020; Groopman 2018; Microsoft 2020					
DaDriDM2	The decisions that are made in my organization are based on data-driven insights.						
Pacouroact	Considering AI's inherent complexity, organizations need dedicated resources to steer the developm	nent of related assets, capabilities, and commitment. Thus, the					
<u>Resources.</u>	category <i>resources</i> considers AI-related financial, personnel, and infrastructural resources.						
	<u>Financial budget</u>	• E08, E09, E10					
FinBud1	My organization has allocated financial resources for AI adoption.	• Alsheibani et al. 2019; Kruse et al. 2019; Pringle and Zoller					
FinBud2	My organization is willing to spend financial resources on AI projects with a high amount of risk	2018; Pumplun et al. 2019					
	and uncertainty.						
	<u>Personnel</u>	• E03, E05, E08, E10					
Per1	My organization has employees with AI know-how.	<ul> <li>Alsheibani et al. 2018; Catalyst Fund 2020; Groopman</li> </ul>					
Per2	My organization has AI specialists who have a deep understanding of AI technologies.	2018; Intel 2018; Kruse et al. 2019; Microsoft 2020; PA					
Per3	My organization has business analysts who possess both domain and AI know-how.	Office of Administration 2018; Pumplun et al. 2019					
_	<u>IT infrastructure</u>	• E01, E03, E10, E11, E13, E14, E15					
ItInf1	My organization's IT infrastructure facilitates data availability, data accessibility, and data flow.	<ul> <li>Alsheibani et al. 2018; Catalyst Fund 2020; Groopman</li> </ul>					
ItInf2	My organization's IT infrastructure can provide processing power for AI workloads.	2018; Intel 2018; Kruse et al. 2019					
ItInf3	My organization's IT infrastructure is modular and allows for the integration of new applications.						

Table 5 Overview of AI Readiness Categories, Factors, and Illustrative Indicators

<u>Knowledge</u> :	Since AI raises questions regarding the applicability and explainability of underlying intelligent tec understanding and expectations of employees toward AI	hniques, the category <i>knowledge</i> reflects the adequate
	AI awareness	• E01 E03 E00 E10 E15 E10
AiAwa1	Employees in my organization are aware of AI's opportunities.	• Catalyst Fund 2020: Kruse et al. 2010: Pringle and Zoller
AiAwa2	Employees in my organization are aware of how AI works.	2018
AiAwa3	Employees in my organization have adequate expectations toward AI.	-010
	Upskillina	• E01 E05 E12 E17 E23 E01 E03 E00 E10 E15 E10
Ups1	Employees in my organization have access to a wide range of upskilling programs.	<ul> <li>Groopman 2018: Intel 2018: Microsoft 2020</li> </ul>
Uns2	Employees in my organization are encouraged to learn new skills.	• Groophian 2010, Intel 2010, Microsoft 2020
Ups3	Employees in my organization are willing to take part in upskilling programs.	
- P-0	AI ethics	• F08 F15 F16
AiEth1	My organization has measures to ensure compliant and ethical conduct	• Catalyst Fund 2020: Groopman 2018: Kruse et al. 2010:
AiFth2	My organization acknowledges that AL posses challenges to algorithmic decisions' explainability	Microsoft 2020; PA Office of Administration 2018
AiFtha	My organization acknowledges that AI requires new measures in order to prevent discrimination	Microsoft 2020, 111 Office of Minimistration 2010
7112013	and safety violations	
Culture	The category culture considers creating an environment that facilitates an openness toward innova	tion and change for AI adoption on an organizational and
<u>outuro</u> .	individual level.	tion and onange for the adoption on an organizational and
	Innovativeness	• E03. E11. E12. E13. E16
Inno1	Employees in my organization experiment to improve established assumptions and practices.	• Catalyst Fund 2020: Groopman 2018: Microsoft 2020:
Inno2	Employees in my organization are willing to innovate radically	Pumplun et al. 2010
Inno2	Employees in my organization are willing to innovate ranidly	r umpfull et ul. 2019
iiiioj	Collaborative work	• For F10 F18
CollWo1	Wy organization has formats and tools to foster collaboration between domain experts. AI	Catalyst Fund 2020: Groopman 2018: Microsoft 2020
Control	specialists and IT	• Catalyst Fund 2020, Groopinan 2010, Microsoft 2020
CollWo2	My organization is willing to facilitate intraorganizational collaboration between domain experts	
0011102	Al specialists and IT through new formats and tools.	
	Chanae management	• E02 E07 E08 E10 E16 E24
ChaMa1	My organization manages resistance to change effectively.	• Pumplun et al 2010
ChaMa2	Change champions, multipliers, and consultants facilitate organizational change in my	
Chabiaz	organization	
ChaMa3	My organization is willing to provide resources and guidance to handle AI-induced change.	
Data:	The category data comprises assets, capabilities, and commitment to ensure high data availability.	quality, accessibility, and flow,
<u>D'utu</u>	Data availabilitu	• E03 E05 E08 E10 E11 E12 E13 E22 E24
DAvail1	My organization has extensive amounts of data about resources, processes, transactions, and	<ul> <li>Catalyst Fund 2020: Groopman 2018: Intel 2018: Kruse et</li> </ul>
Diritani	other events related to my organization.	al 2010: Microsoft 2020: Pringle and Zoller 2018.
DAvail2	Data about resources, processes, transactions, and other events related to my organization is	Pumplun et al. 2010
Dirum	digitally available instead of paper-based.	
DAvail3	Data within my organization is available in the form of structured data.	
DAvail4	My organization has methods and procedures to transform unstructured or semi-structured data	
Dirtunq	into structured data.	
DAvail5	My organization has data that is relevant for potential AI use cases.	
DAvail6	My organization regularly identifies and establishes new sources to collect data.	
	Data quality	• E08, E11, E13, E15, E24
DaOua1	Data within my organization represents real-world events.	• Catalyst Fund 2020: Groopman 2018: Kruse et al 2010:
DaQua2	Data within my organization is correct.	PA Office of Administration 2018: Pringle and Zoller 2018.
DaQuaz	Data within my organization is complete	Pumplun et al. 2010
Duquuj	Data manning organization is complete.	

DaQua4	My organization has methods and procedures for anomaly detection.	
DaQua5	My organization has methods and procedures for data cleaning.	
DaQua6	My organization has methods and procedures for data quality assurance.	
	<u>Data accessibility</u>	<ul> <li>E03, E08, E12, E13</li> </ul>
DaAcc1	Data within my organization is quickly and easily retrievable.	<ul> <li>Catalyst Fund 2020; Groopman 2018; Intel 2018</li> </ul>
DaAcc2	Data within my organization is accessible through centralized data sources.	
DaAcc3	Data within my organization is accessible across different departments.	
DaAcc4	Ways to access data within my organization are documented.	
	<u>Data flow</u>	• E08, E10, E13, E24
DaFlo1	My organization moves data from its source to its use in an automated manner.	<ul> <li>Catalyst Fund 2020; Groopman 2018</li> </ul>
DaFlo2	In my organization, moving data from its source to its use does not produce data corruption.	
DaFlo3	In my organization, technological bottlenecks do not hamper moving data from its source to its	
	use.	



#### Appendix V: Exemplary AI Readiness Assessment

Fig. 4 Exemplary AI Readiness Assessment