







Session 7: AI readiness & service platforms

Managing Al-based Systems

Prof. Dr. Nils Urbach

Frankfurt University of Applied Sciences, Research Lab for Digital Innovation & Transformation

FIM Forschungsinstitut für Informationsmanagement

Fraunhofer-Institut für Angewandte Informationstechnik FIT, Institutsteil Wirtschaftsinformatik

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Course navigator



Basic understanding of AI and its business potential

Al Ideation

Application potentials of Al technologies

Identification, design and evaluation of AI use cases

Al Strategizing

Evaluation an organization's Al readiness

Management and governance of Al

Al Design & Development

Architectures of Al applications

Data Management and Model Transparency

Design of human-Al interaction

Al Operations at Scale

Monitoring and KPIbased control

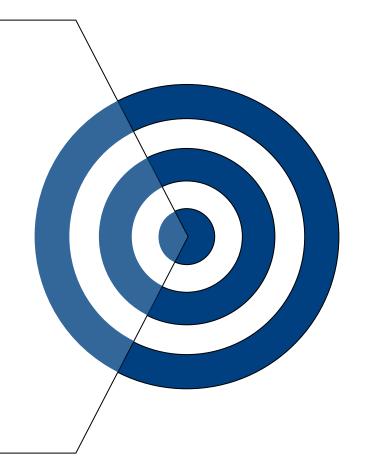
Ethical, legal and social implications of Al

Implementation

Objectives of today's lecture

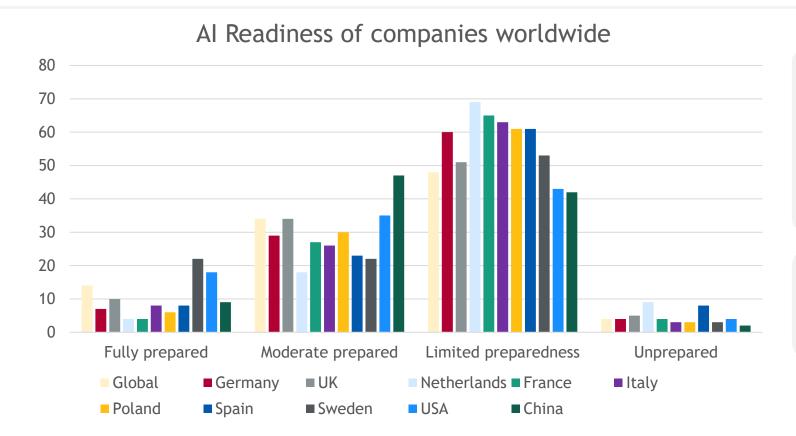


- 1. You understand why it is important to carefully to assess AI readiness
- 2. You will be able to evaluate the AI readiness of a company
- 3. You get to know different specifications to choose a fitting AI service platform



Many companies feel not to be well-prepared for AI





Sub-groups of AI readiness

- Germany lacks especially in infrastructure and data readiness compared to the global average
- Regarding AI Strategizing, German companies feel quite confident and nearly reach global average

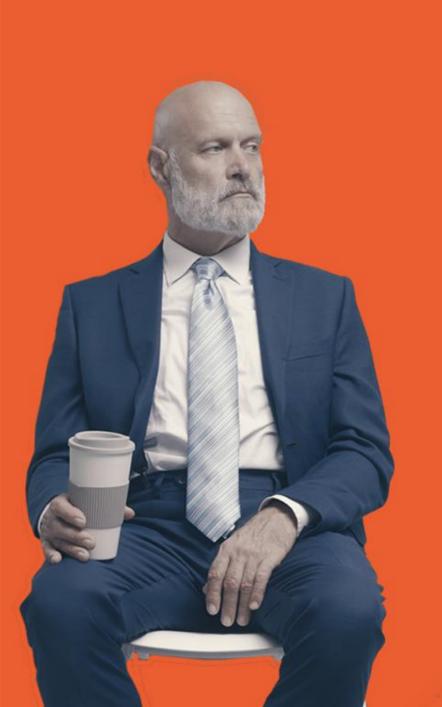
Information on sample size

- 8161 Executives of companies (>500 empl.)
- 30 countries
 - 9 European countries



The evaluation of the current data situation shows that German companies are lagging in both European and global comparison. The global average shows that more than every 2nd company in the world does not feel AI ready (sum of limited preparedness and unpreparedness).

Source: Cisco (2023)

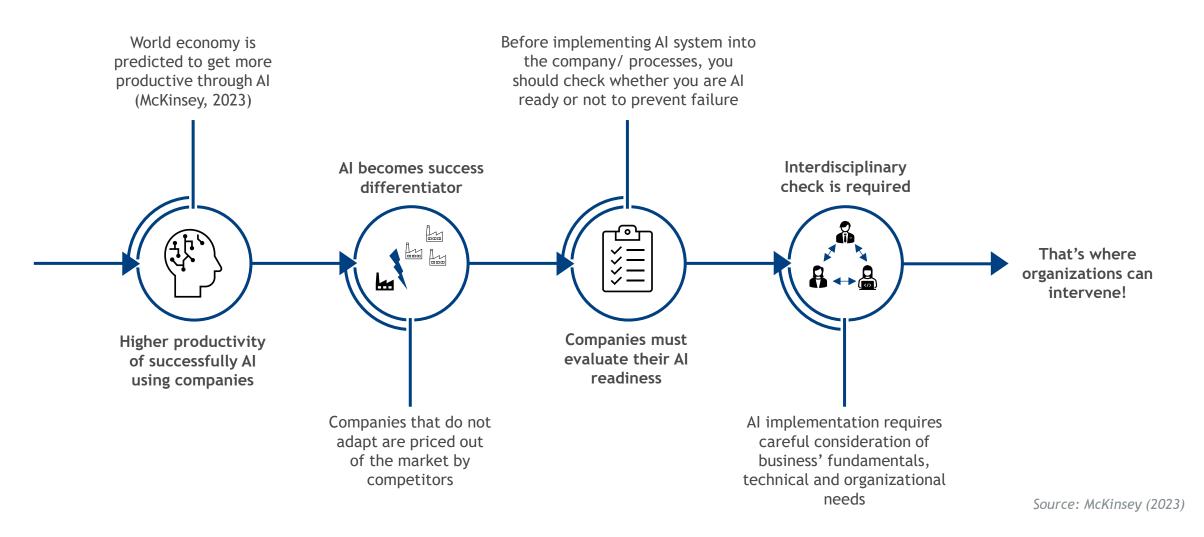






Why is AI readiness important to evaluate?







Al readiness definition and related terms

Organizational AI readiness factors

1 Integrating Al service platforms



Al readiness definition and related terms

Organizational AI readiness factors

1 Integrating Al service platforms



Definition and related terms: AI readiness



Al readiness

The preparedness of organizations to implement change involving applications and technology related to AI (1)

Innovation adoption

(AI is technological innovation)

Innovation adoption refers to the decision to use innovation (expectation of improved organizational performance) and understand the necessity of its organizational use (2)

related to

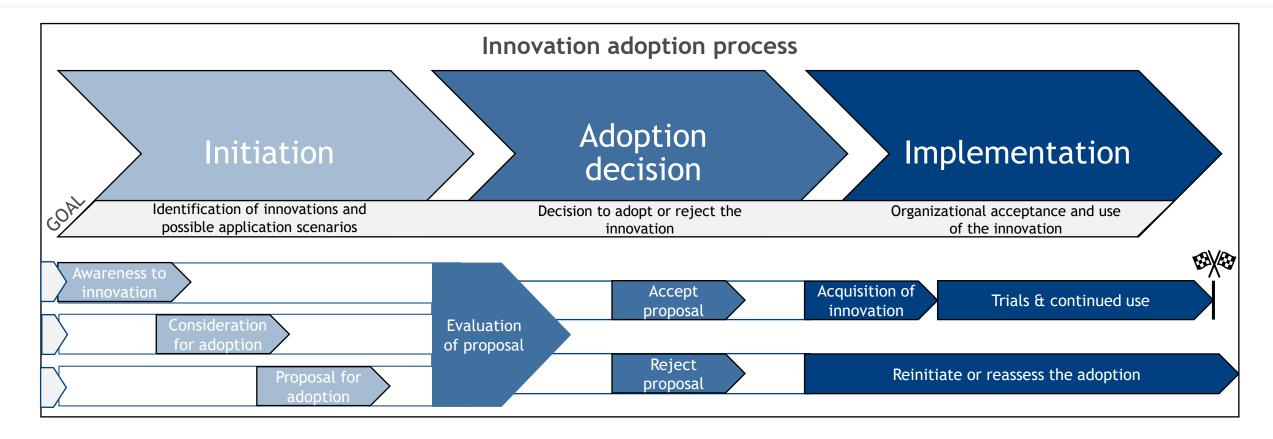
Organizational readiness for change

Readiness indicates the state necessary to engage in a specific activity, e.g., the adoption of a specific innovation such as Al (2,3)

(1) Alsheibani et al. (2018), (2) Jöhnk et al. (2021), (3) Lokuge et al. (2018))

Definition and related terms: Innovation adoption process







The general process of innovation adoption can also be applied for AI implementation. An AI readiness check could be positioned in the step of initiation, especially evaluation of proposal.

Rogers (2003), Gopalakrishnan and Damanpour (1997), Hameed et al. (2012), Jöhnk et al. (2021)

Definition and related terms: Organizational readiness for change



Readiness for technology adoption

- Financial and techn. resources
- Management support
- Organizational culture
- Commitment
- Communication of goals
- Partnership readiness

Readiness for digitalization

- Organizational assets
- Capabilities
- Commitment

Readiness for Al?

Are there specific factors for readiness in artificial intelligence?

Chwelos et al. 2001; Damanpour and Schneider 2006; Iacovou et al. 1995; Lokuge et al. 2018; Nguyen et al. 2019, Robey et al. 2008



01

Al readiness definition and related terms

Organizational AI readiness factors

03

Integrating AI service platforms

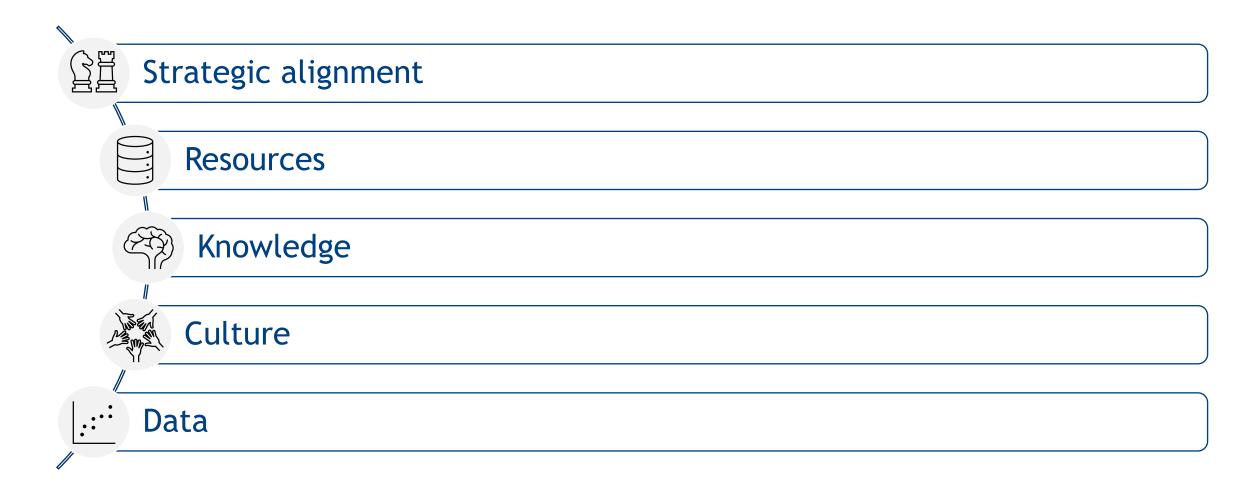


Which factors do you think are important for AI readiness?



Organizational AI readiness factors

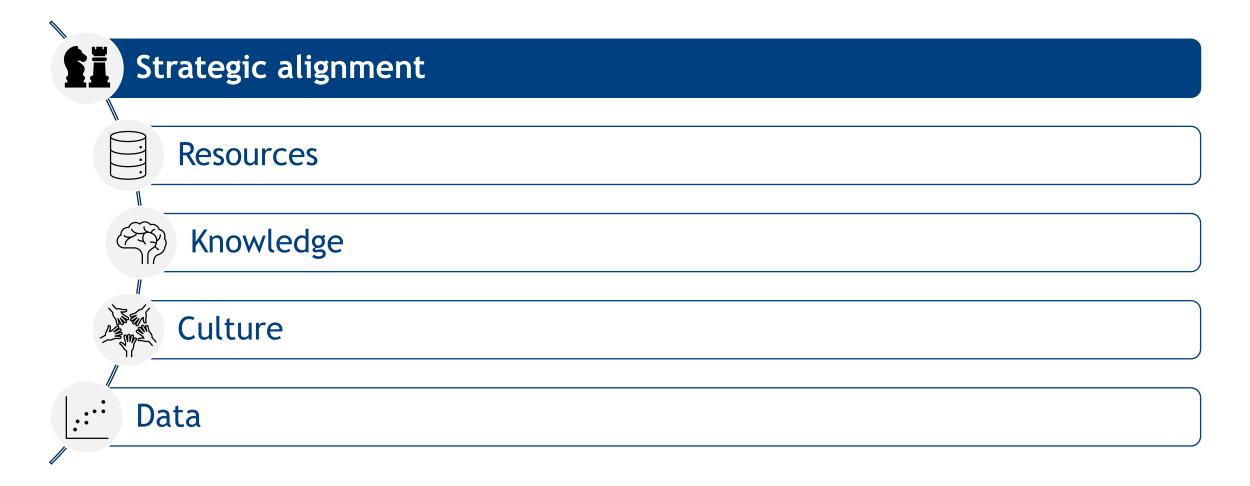




(1) Jöhnk et al. 2021



Organizational AI readiness factors: Strategic alignment



(1) Jöhnk et al. 2021

Organizational AI readiness factors: Strategic alignment





Strategic alignment

Al-business potentials

Al characteristics: Functions are highly versatile and broadly applicable

Organizational necessity: Al-business potentials ensure that Al adoption is beneficial and suitable for organization

Customer Al readiness

Al characteristics: Use requires understanding of complexity, lack of algorithm transparency

Organizational necessity: Customer Al readiness enables in- & external customers to use Al-integrated offerings

Top management support

Al characteristics: Al's complexity poses change not only within but across organizational levels, which requires top management commitment

Organizational necessity: Top management support shows strategic relevance to organization

Al-process fit

Al characteristics: Al-based systems are more precise if processes are structured and provide standardized data input

Organizational necessity: Al-process fit through standardization, reengineering and implementation of new processes facilitates Al adoption

Data-driven decision-making

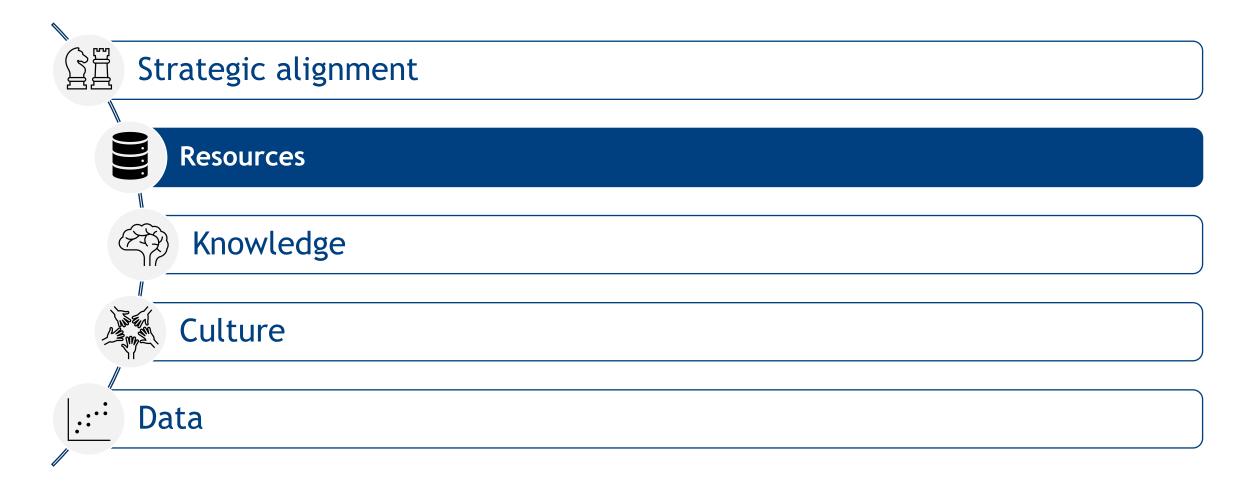
Al characteristics: Al-based systems are data-driven and require openness to incorporate such insights

Organizational necessity: Data-driven decision-making fosters AI adoption because both utilize data and statistical methods to gain insights

Jöhnk et al. (2021)

Organizational AI readiness factors





Organizational AI readiness factors: Resources





Financial budget

Al characteristics: Al-based systems require high investments to tailor assets and capabilities to the unique data

Organizational necessity: Strategic allocation of financial budget for AI adoption supports the overcoming of initial obstacles and uncertainty

Resources

Personal

Al characteristics: Adoption requires a broader spectrum of different roles and know-how for core business use

Organizational necessity: Al specialists and business analysts with specific knowhow facilitate adoption

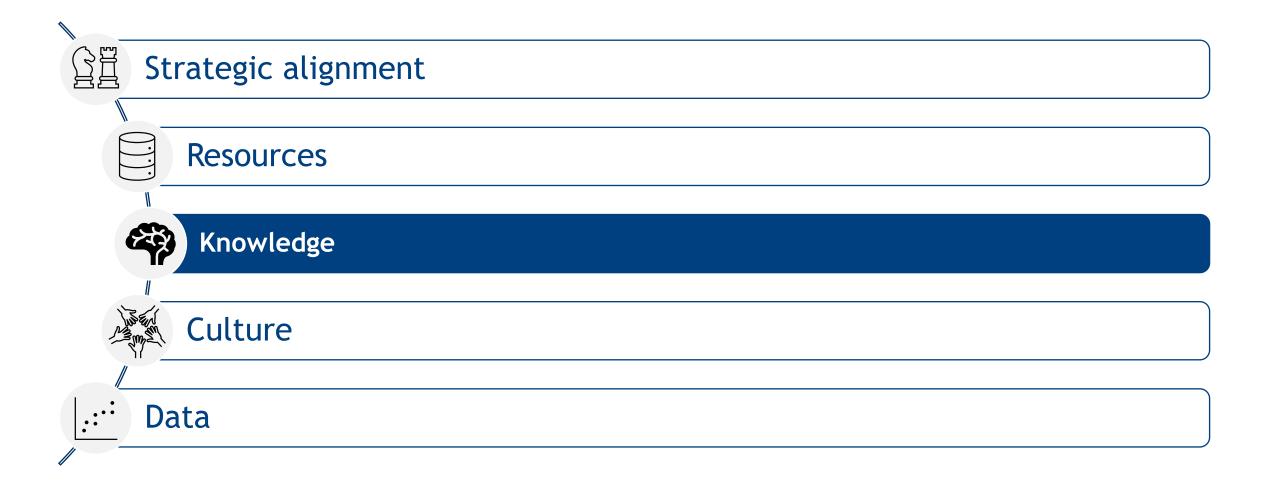
IT infrastructure

Al characteristics: Deploying Al poses high workloads and data storage requirements

Organizational necessity: IT infrastructure enables AI-related activities and AI integration

Organizational AI readiness factors





(1) Jöhnk et al. 2021

Organizational AI readiness factors: Knowledge





Knowledge

Al awareness

Al characteristics: Al's underlying concepts (e.g., machine learning) are hard to grasp

Organizational necessity: Al awareness ensures that employees have adequate understanding and expectations toward Al

Upskilling

Al characteristics: Requires every employee to have a basic understanding of Al

Organizational necessity: Upskilling enables employees to learn and develop AI(-related) skills

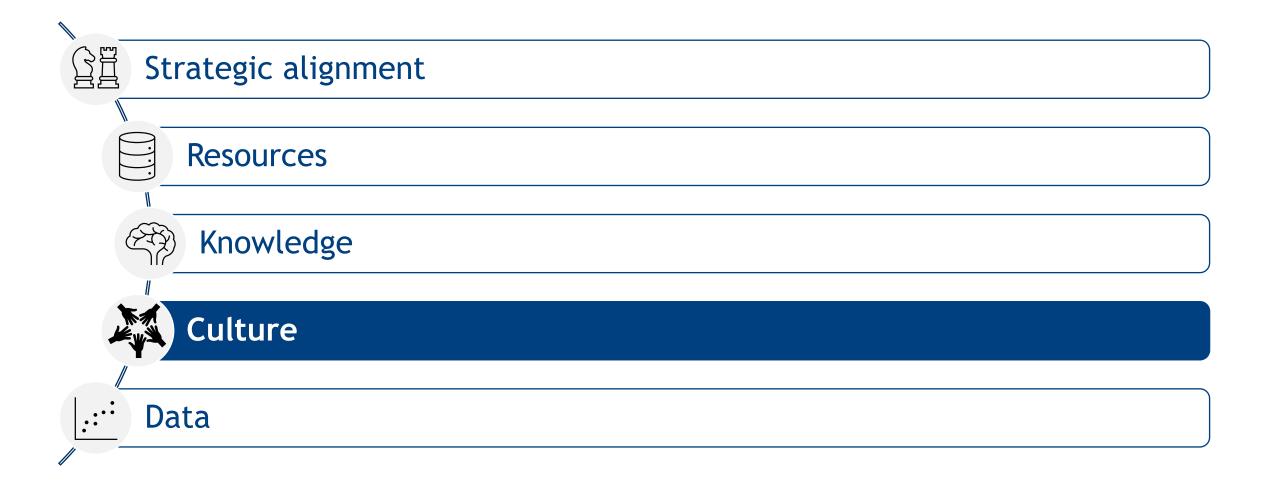
Al ethics

Al characteristics: Al-based systems are at risk for biased learning and unethical outcomes

Organizational necessity: All ethics comprise measures to prevent bias, safety violations or discrimination in All outcomes

Organizational AI readiness factors





(1) Jöhnk et al. 2021

Organizational AI readiness factors: Culture





Innovativeness

Al characteristics: Employees' fear of Al-induced job loss threatens proactive innovativeness

Organizational necessity:

Innovativeness increases employees' willingness to change the status quo through the application of AI

Culture

Collaborative work

Al characteristics: Al deployment relies on integrating different perspectives (i.e., domain, data and IT)

Organizational necessity: Collaborative work enables employees to work in teams and combine different skills

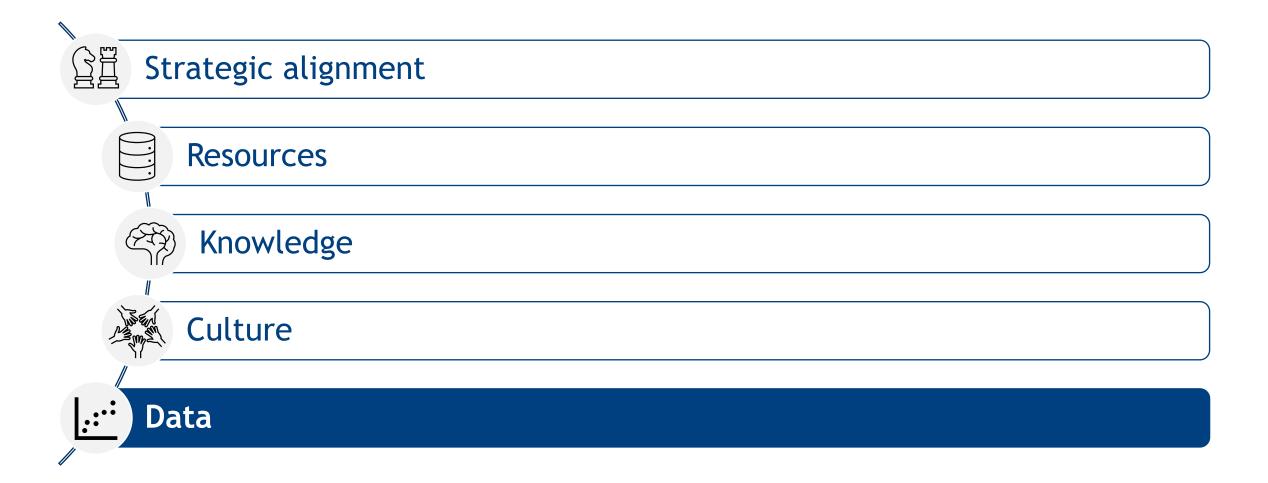
Change management

Al characteristics: Employees' lack of understanding and fear of Al threaten the acceptance of Al-based systems

Organizational necessity: Change management helps employees to understand and cope with AI-induced organizational change

Organizational AI readiness factors





(1) Jöhnk et al. 2021

Organizational AI readiness factors: Data





Data availability

Al characteristics: Al-based systems learn through different data types and large data amounts

Organizational necessity: Data availability within the organization fuels Al solutions

Data

Data quality

Al characteristics: Al-based systems achieve better results the higher the quality of the data, they learn with, is

Organizational necessity: Data quality ensures accurate Al outcomes

Data flow

Al characteristics: Initial and continuous training of Al-based systems requires smooth and automated data flow

Organizational necessity: Data flow between its source and its use ensures high data accessibility to AI experts

Data accessibility

Al characteristics: Al personnel require access to relevant data sources for deployment

Organizational necessity: Data accessibility facilitates AI experts to easily prototype and develop AI solutions

Jöhnk et al. (2021)

Recommended reading



Bus Inf Syst Eng 63(1):5-20 (2021) https://doi.org/10.1007/s12599-020-00676-7



RESEARCH PAPER

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

Jan Jöhnk · Malte Weißert · Katrin Wyrtki

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Abstract Artificial intelligence (AI) offers organizations much potential. Considering the manifold application areas, AI's inherent complexity, and new organizational necessities, companies encounter pitfalls when adopting AI. An informed decision regarding an organization's readiness increases the probability of successful AI adoption and is important to successfully leverage AI's business value. Thus, companies need to assess whether their assets, capabilities, and commitment are ready for the individual AI adoption purpose. Research on AI readiness and AI adoption is still in its infancy. Consequently, researchers and practitioners lack guidance on the adoption of AI. The 1 Introduction paper presents five categories of AI readiness factors and their illustrative actionable indicators. The AI readiness factors are deduced from an in-depth interview study with (GPT) with a unique learning capability that provides 25 AI experts and triangulated with both scientific and practitioner literature. Thus, the paper provides a sound set of organizational AI readiness factors, derives

Electronic Supplementary Material The online version of this plementary material, which is available to authorized users.

Accepted after two revisions by the editors of the special issue

J. Jöhnk () · K. Wyrtki Project Group Business & Information Systems Engineering of the Fraunhofer FIT, FIM Research Center, Wittelsbacherring 10, 95444 Bayreuth, Germany

e-mail: jan.joehnk@fit.fraunhofer.de e-mail: katrin.wvrtki@fit.fraunhofer.de

FIM Research Center, University of Bayreuth, Universitätsstraße 30, 95447 Bayreuth, Germany e-mail: malteweissert@gmail.com

corresponding indicators for AI readiness assessments, and discusses the general implications for AI adoption. This is a first step toward conceptualizing relevant organizational AI readiness factors and guiding purposeful decisions in the entire AI adoption process for both research and practice.

Keywords Artificial intelligence · AI adoption · AI readiness · Organizational readiness assessment · Interview

Artificial intelligence (AI) is a general-purpose technology organizations with potentials for wide-ranging improvements as well as entirely new business opportunities. Iansiti and Lakhani 2020. Being a GPT, AI drives changes at the task, process, and business model level in a plethora of application areas and as such offers a competitive advantage to organizations (Brynjolfsson and McAfee 2017). Ample access to improved learning algorithms and available AI use cases as well as corresponding solutions facilitate AI adoption in organizations. In 2019, 80% of large organizations aimed to adopt or had adopted some form of AI (Gartner 2019; Ghosh et al. 2019). However, most organizations use AI in single pilots, whereas only 8% of organizations have adopted AI in core practices (Fountaine et al. 2019). Due to Al's nature as a GPT, possible application scenarios are not always directly obvious and organizations must understand the technology to decide on the intended adoption purpose (Jovanovic and Rousseau 2005). Consequently, the adoption of AI poses challenges on an organizational, technical, and individual level (Baier et al. 2019; Bughin et al. 2017). Thus,

Springer

Jöhnk, J., Weißert, M., & Wyrtki, K. (2021) Ready or Not, Al Comes — An Interview Study of Organizational Al Readiness Factors, Business & Information Systems Engineering, 63, 5-20.





01

Al readiness definition and related terms

02

Organizational AI readiness factors

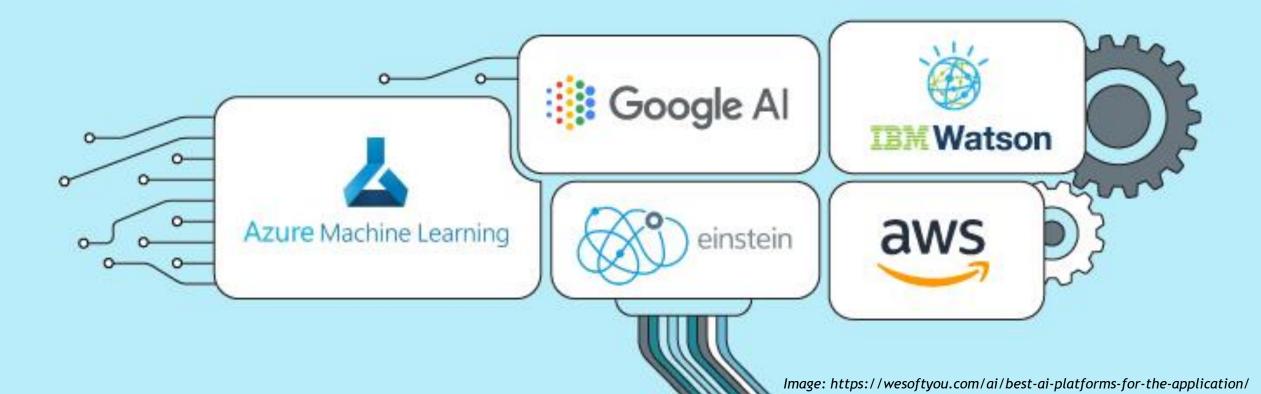
03

Integrating AI service platforms

Al platforms have evolved to cater to meet the diverse needs of different industries

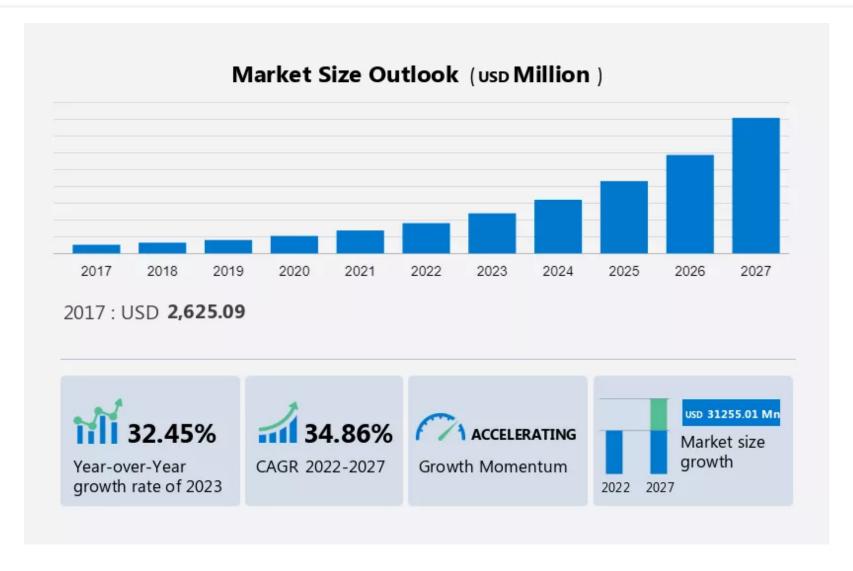


TOP AI PLATFORMS ACROSS INDUSTRIES



The artificial intelligence platforms market is estimated to grow at a CAGR of 34.86% between 2022 and 2027





Source: Technavio (2023)

Examples of AI service platforms: Microsoft Azure



Azure Machine Learning

Azure ML is an integrated, comprehensive data science and advanced analytics solution; it offers data preparation, collaborative notebooks and automated machine learning

Application example

A healthcare company could use Azure ML to analyze patient data and make predictions about future health risks

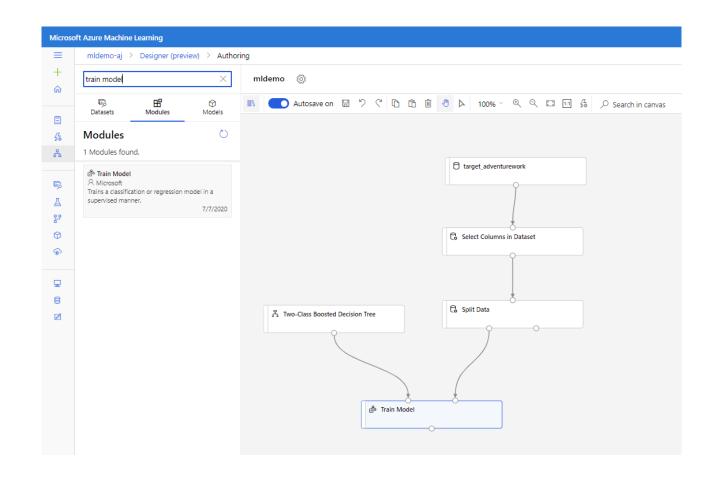


Image: Microsoft

Examples of AI service platforms: AWS

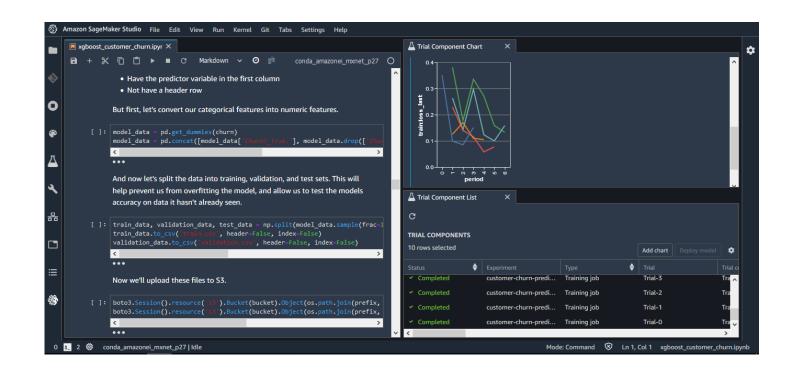


Amazon Web Services (AWS) Machine Learning

AWS offers a variety of machine learning and deep learning services, including Amazon SageMaker for training and deploying models

Application example

Organizations can use SageMaker to create, train and deploy custom models without having to worry about the underlying infrastructure



Examples of AI service platforms: Google Cloud



Google Cloud (Al and Machine Learning)

Google Cloud offers services for generative AI, machine learning, speech processing, document and data, vision and video and AI infrastructure

Application example

An e-commerce company can use Vision AI to analyze images in its catalogue and generate product recommendations

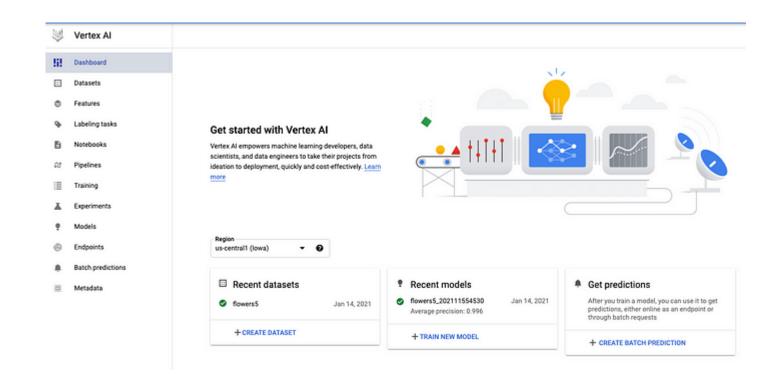


Image: Towards data science

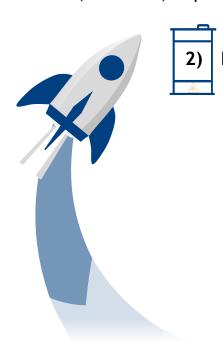






High degree of freedom to create your own AI models or applications:

- a. Platform enables organizations to create their own AI models with a high degree of freedom
- b. Provides generic AI development services
- c. End-to-end services with various benefits for customers and providers (ease of use, operating costs, (technical) dependencies, and reducing complexity)



Ready-to-use AI applications:

- a. Platform offers integrated AI functionalities
- b. Limited development effort for customers
- c. Continuous spectrum between development platforms and finished AI services

3)

Focus on secondary Al services and resources:

- a. Platform offers complementary services such as data labeling and ML model assessment
- b. Access to pre-built ML models and resources
- c. Support for the development process of AI solutions

Geske et al. (2021)

How to choose a fitting AI service platform



Al Service Platform - Characteristics

Customer context

Locates and describes the area of application as well as assesses the AI service platform's specialization from a customer's perspective

Al service platform's offering

Assesses the value proposition of the platform

Third party integration

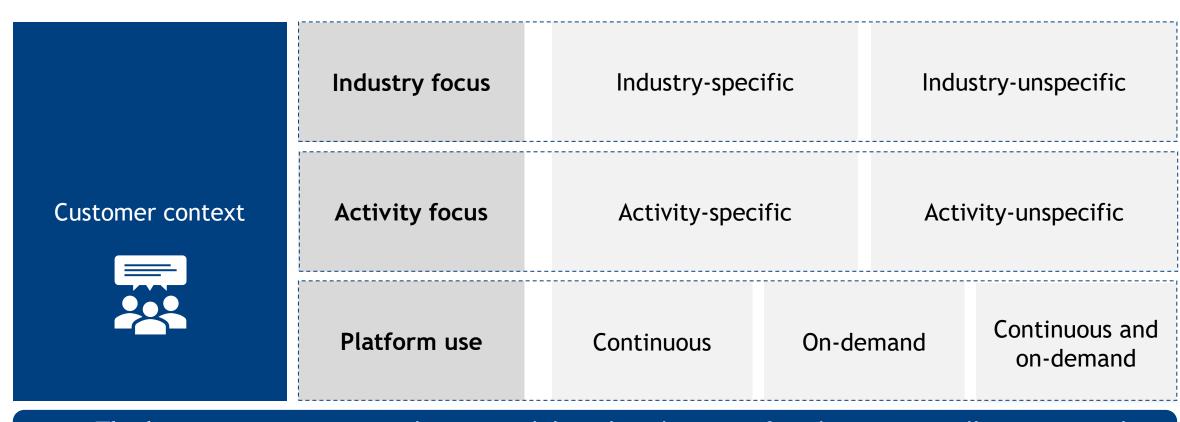
Describes the platform's constitutive agents beyond the platform provider and the customer



The top three layers that help to define the Al Service platform are the customer context, the Al service platform's offering and the third-party integration

Specifying the AI service platform: Customer context





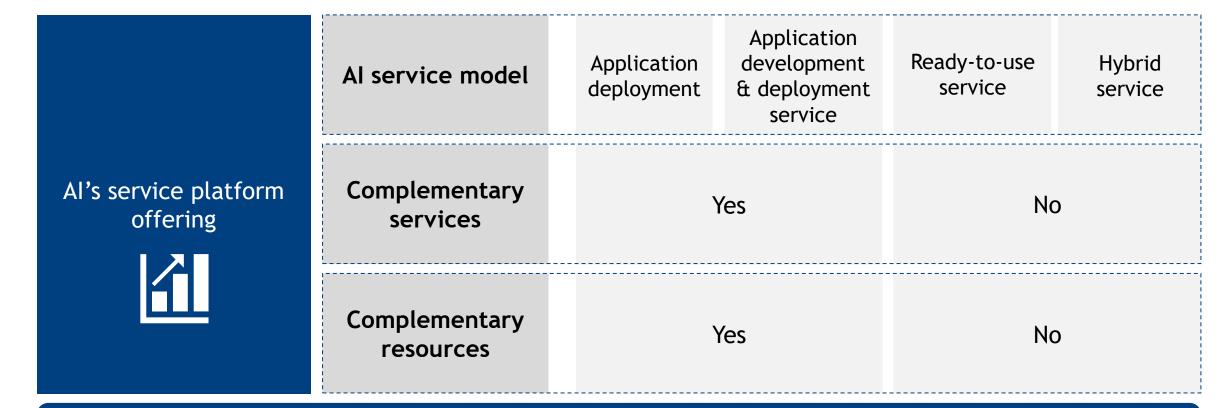


The layer customer context locates and describes the area of application as well as assesses the AI service platform's specialization from a customer's perspective.

Gawer (2014), Geske et al. (2021), James and Chung (2015)

Specifying the AI service platform: Platform offering





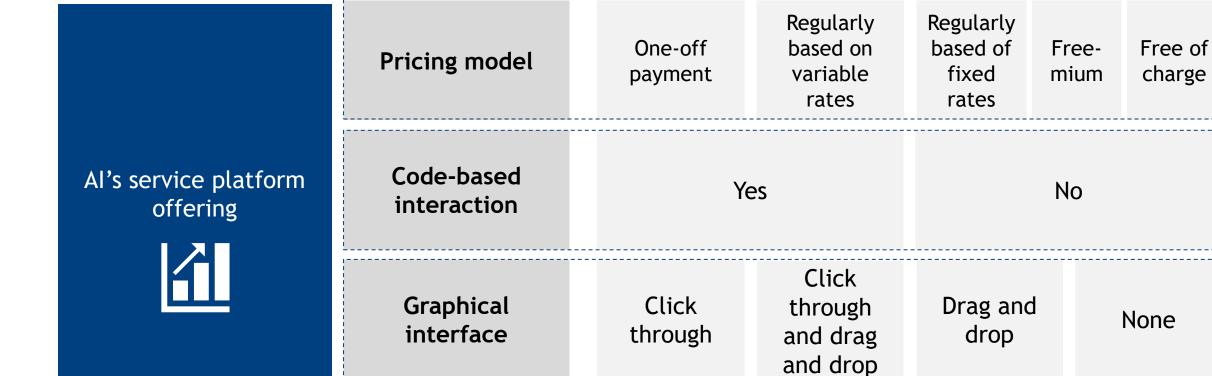
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The layer Al's service platform offering assesses the value proposition of the platform

Gawer (2014), Geske et al. (2021)

Specifying the AI service platform: Platform offering





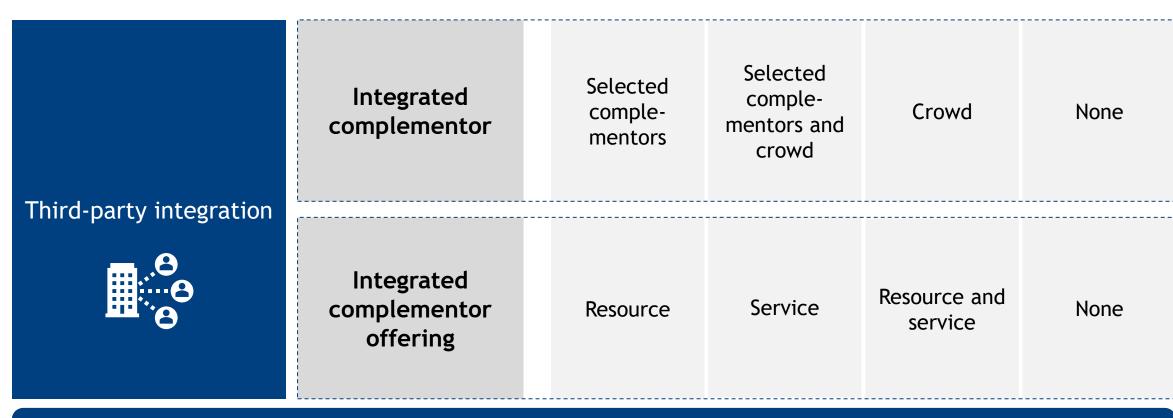
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The layer Al's service platform offering assesses the value proposition of the platform

Gawer (2014), Geske et al. (2021)

Specifying the AI service platform: Third-party integration







The layer third-party integration describes the platform's constitutive agents beyond the platform provider and the customer

Gawer (2014), Geske et al. (2021)

Recommended reading



AI Service Platform Taxonomy

GATEWAYS TO ARTIFICIAL INTELLIGENCE: DEVELOPING A TAXONOMY FOR AI SERVICE PLATFORMS

Research Paper

Flora Geske, Technical University of Munich, Munich, Germany, flora.geske@tum.de

Peter Hofmann, Project Group Business & Information Systems Engineering of the Fraunhofer FIT, FIM Research Center, University of Bayreuth, Bayreuth, Germany, peter hofmann@fit fraunhofer.de

Luis Lämmermann, Project Group Business & Information Systems Engineering of the Fraunhofer FIT, FIM Research Center, University of Bayreuth, Bayreuth, Germany, luis Jaemmermann@uni-bayreuth.de

Vincent Schlatt, Project Group Business & Information Systems Engineering of the Fraunhofer FIT, FIM Research Center, University of Bayreuth, Bayreuth, Germany, vincent.schlatt@fit.fraunhofer.de

Nils Urbach, Frankfurt University of Applied Sciences, Frankfurt, Germany, Project Group Business & Information Systems Engineering of the Fraunhofer FIT, Bayreuth, Germany, nils urbach@fim-rc.de

Abstract

Artificial Intelligence (AI) carries the potential to drive innovation in many parts of today's business environment. Instead of building AI capabilities in-house, some organizations turn towards an emergent phenomenon: AI service platforms. However, as a novel concept in both research and practice, a systematic characterization of AI service platforms is missing. To address this gap, we define the concept of AI service platforms and develop a comprehensive atxonomy. Therefore, we rely on existing literature, 14 expert interviews, and a sample of 31 AI service platforms. Our contribution is threefold: First, our taxonomy systematically structures essential properties of AI service platforms, guiding future research and management practice. Second, we derive three generic motives of AI service platforms if into the existing academic discourse on digital platforms and elaborate on fluture research directions.

Keywords: AI Service Platform, Digital Platform, Artificial Intelligence, Taxonomy.

1 Introduction

Fostering developments in new products, services, and business models, Artificial Intelligence (AI) is experiencing attention in research and practice. AI applications are diverse and involve different technological approaches. Examples range from classification in medical diagnosis (Hofmann et al., 2019) to the generation of music (Hofmann et al., 2021). Today, AI has moved from the theoretical realm to the global marketplace. Amongst others, the maturing of machine learning (ML), the availability of data, and adequate computing power have fueled the rise of AI (Jordan and Mitchell, 2015). Despite AI's promised potential, companies have not yet widely adopted AI. Lorica and Paco (2019) state that only 27 percent of the observed companies are already deploying AI, while another 34

Twenty-Ninth European Conference on Information Systems (ECIS 2021), [Marrakesh, Morocco]A Virtual AIS Conference].

Geske, F., Hofmann, P., Lämmermann, L., Schlatt, V. and Urbach, N. (2021) Gateways to Artificial Intelligence: Developing a Taxonomy for Al Service Platforms, Proceedings of the 29th European Conference on Information Systems (ECIS 2021), June 14-16, Marrakesh, Marocco.



Today's lecture at a glance



We discussed different factors to evaluate the AI readiness of a company

We understand when companies should build their own platform and when they should approach an AI service platform

We are able to choose the right AI service platform by choosing the right specifications

Questions, comments, observations





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Pictures



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