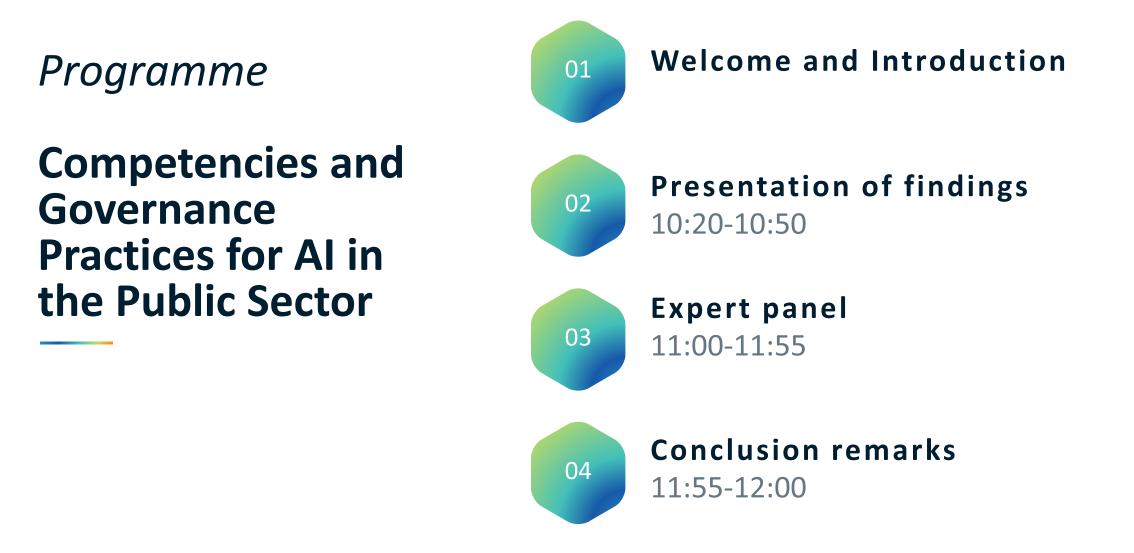
Public Sector Tech Watch

Webinar Series: "Innovating the EU public sector in Europe"

1st Webinar Competencies and Governance Practices for AI in the Public Sector

23 April 2024







Competencies & Governance Practices for AI in the Public Sector

An empirical study

Methodology

1/3

Literature review

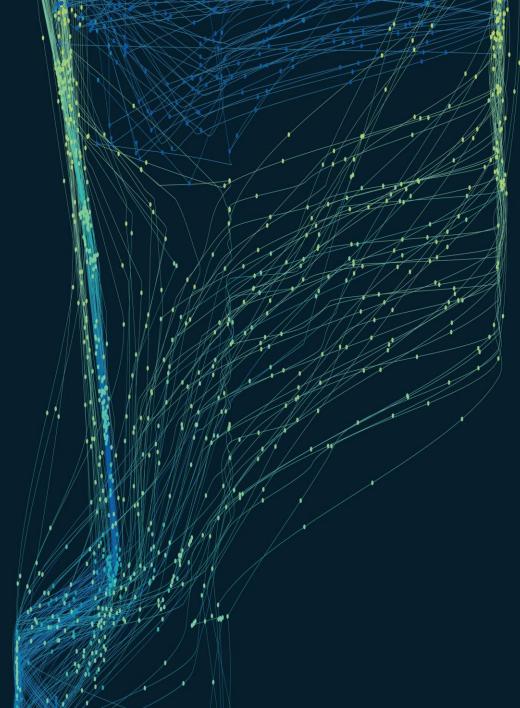
- As of December 2023
- Academic literature (keyword search)
- Policy / grey literature (snowball approach)



	Competencies	Governance	Total unique
Academic	18	15	30
Policy	14	5	18
Total	32	20	48

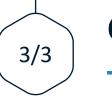






Methodology

Methodology



Case studies

- 7 countries
- May-November 2023
- 19 semi-structured interviews
- 11h 36m recordings



Country	Organization	
Czechia	Ministry of Interior	
Denmark	Municipality of Gladsaxe	
Germany	District of Lüneburg	
Greece	Ministry of Digital Governance	
Italy	National Institute for Social Security	
Norway	Municipality of Trondheim	
The Netherlands	City of Amsterdam	



Al competencies in the public sector



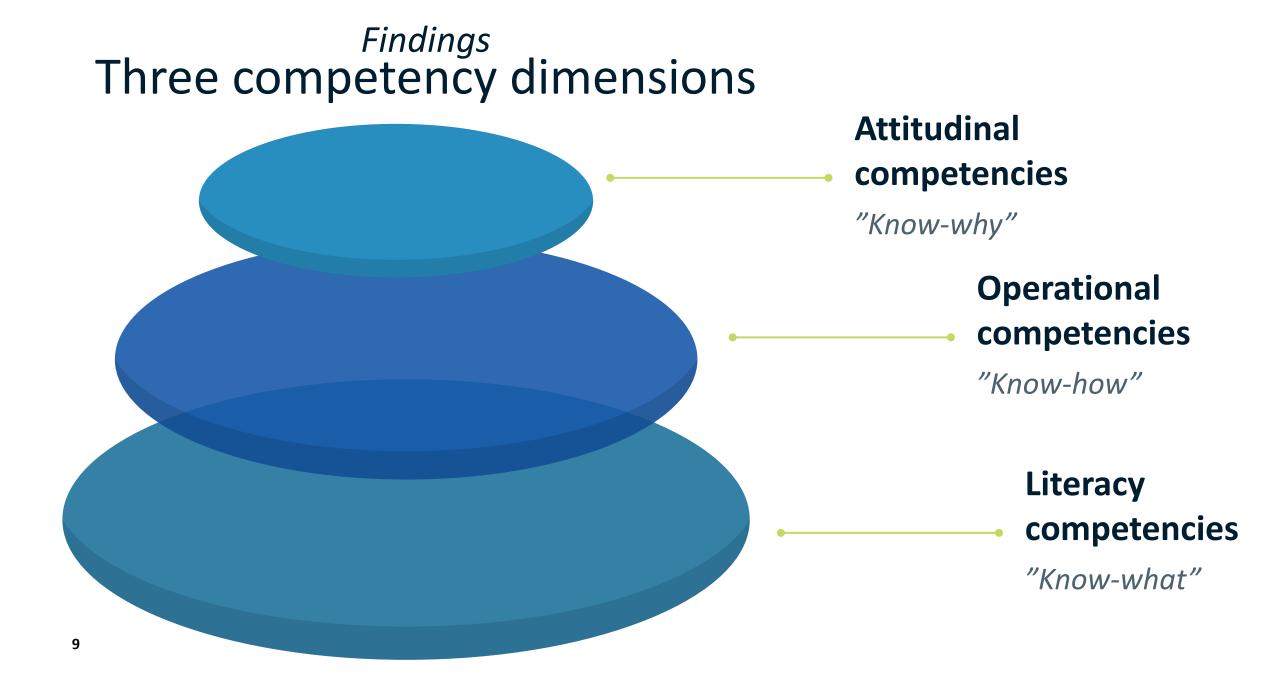
Public managers often have **limited expertise**, and lack **competent figures** to harness the benefits of AI technologies (Ahn & Chen, 2022; Neumann et al., 2022)

Definition

Knowledge, skills and behaviors that are visible in an individual; and individual underlying attributes such as traits, motives, attitudes, values and self-image that tend to be deeper (Salman et al., 2020)

Findings Three competency areas





Technology	Attitudinal competencies (<i>know-why</i>)	 Technology inquisitiveness Positive attitude towards AI Technical design thinking Data-oriented culture
	Operational competencies (<i>know-how</i>)	 Database management Data governance Data collection Data modelling Data quality assessment Data collection Data collection Data collection Data sharing Choice of AI architecture Choice of ML techniques AI-related software programming Algorithm training Compliance with AI technical standards Prompt engineering
10	Literacy competencies (<i>know-what</i>)	 Basic data literacy Understanding of causal analysis and decision theory Understanding the fundamentals of ML Understanding of AI computer vision

Technology	Attitudinal competencies (<i>know-why</i>)	CR You need to understand how and why it [a Large Language model] is answering wrong.
	Operational competencies	 Database mana Data governance Data collection
	(know-how)	 Data modelling Data quality assessment No: it's just the predicting of the next word. It hasn't got a mind.
11	Literacy competencies (<i>know-what</i>)	 Basic data literacy Understanding of causal analysis and otheory Understanding the fundamentals of NLP Understanding of AI computer vision

Hanagerial	Attitudinal competencies (<i>know-why</i>)	 Leadership Foresight Risk-proclivity Al benefits understanding User-centricity User-centricity Multidisciplinarity Project ownership
	Operational competencies (<i>know-how</i>)	 Risk anticipation and mitigation Choice to delegate to AI Knowledge brokering Cross-team collaboration Data-supported decision-making Coordination Coordination Inter-group translation Partnership development Change management
12	Literacy competencies (<i>know-what</i>)	

Hanagerial	Attitudinal competencies (<i>know-why</i>)	 Leadership Foresight Risk-proclivity Al benefits und 	We need some people that can understand the organization's problems and wants.
	Operational	 Risk anticipation and Choice to delegate to 	The technicians are not very good at that.
	competencies (<i>know-how</i>)	 Knowledge brokering Cross-team collaboration Data-supported decision-max 	They need to go into a dialogue with the departments.
13	Literacy competencies (<i>know-what</i>)		

Policy Legal (know-why) • Awareness of		 Empathy Critical technology assessment Awareness of sustainability implications Policy design thinking
	Operational competencies (<i>know-how</i>)	 Al-compatible policy formulation Auditing Dissemination Collaboration with domain experts Collaboration with Al ethicists
14	Literacy competencies (<i>know-what</i>)	 Al procurement literacy Understanding of legal and ethical frameworks Understanding of public policy making and theory Specialised legal expertise Privacy and security literacy Awareness of ethical implications



Attitudinal competencies (*know-why*)

Operational	٠	Al-compati
competencies	•	Auditing 🛛 🏹
(know-how)	•	Dissemination

Being critical towards these [IT vendor] companies,

understanding what their interests are,

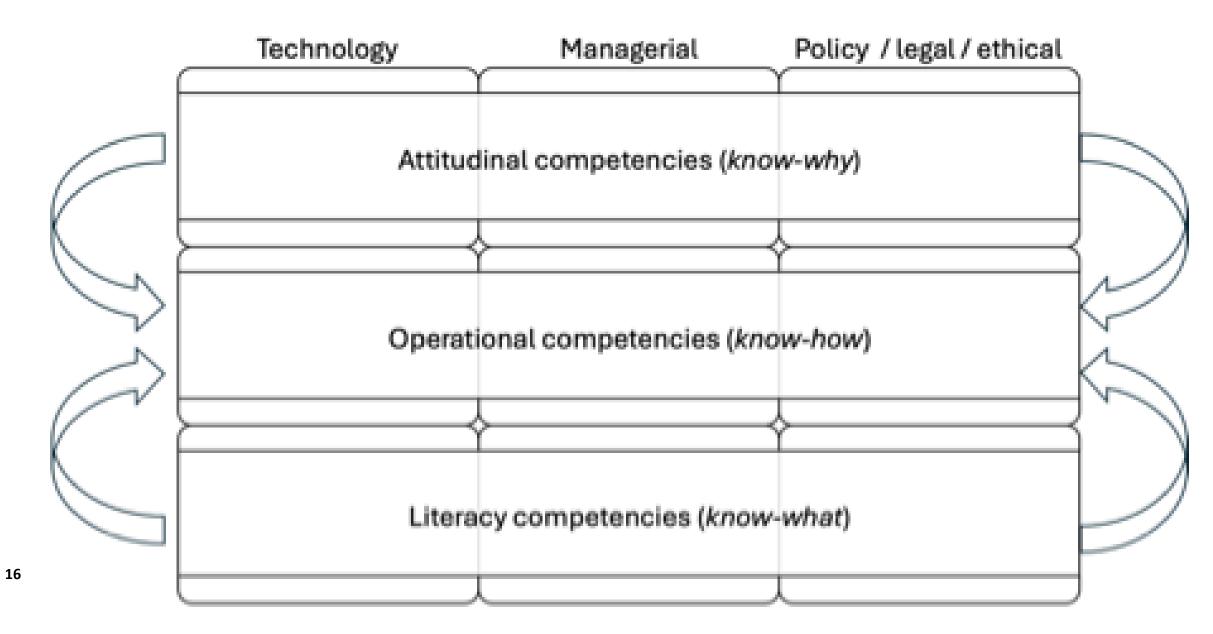
being able to actually technically compare what they're offering to someone else.

Literacy competencies (*know-what*)

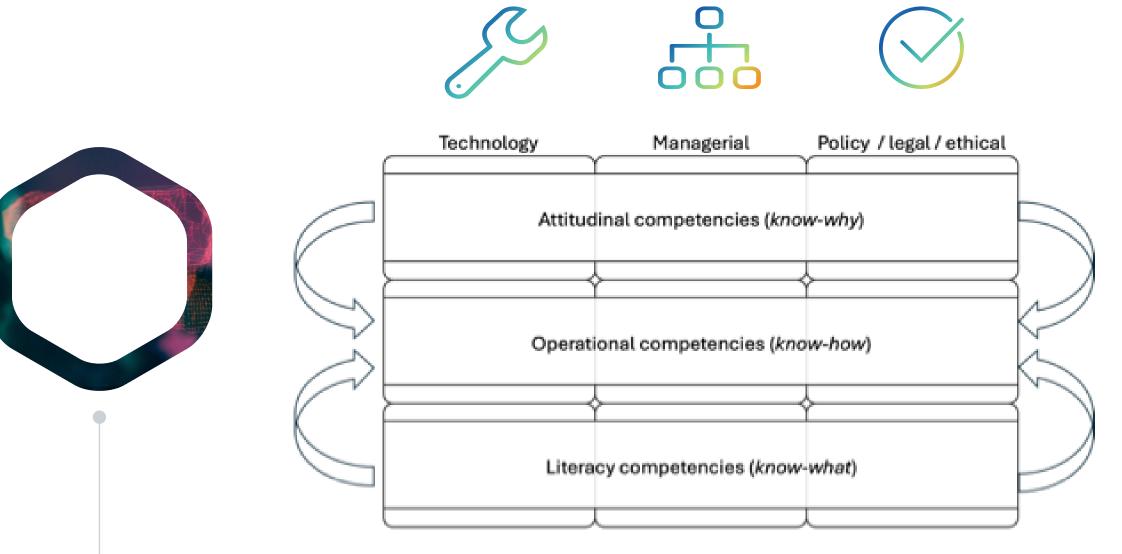
Al procurement literacy

- Understanding of legal and ethical
- Understanding of public policy making a theory

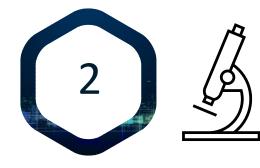
Competencies for AI in the public sector: a comprehensive framework



Recommendations







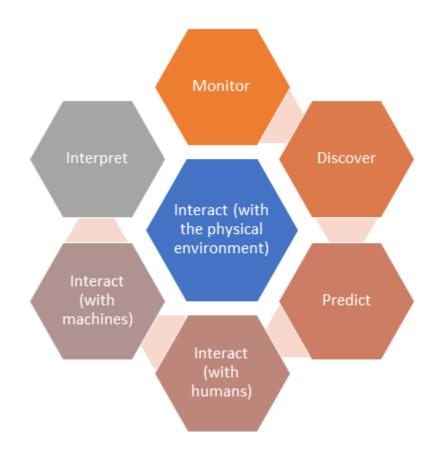
PROMOTE INTERDISCIPLINARY RESEARCH ON AI COMPETENCIES



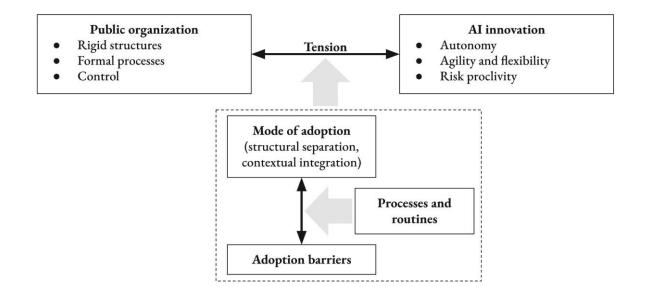
ESTABLISH DEDICATED HIRING PROCESSES and DEVOTE ADDITIONAL RESOURCES TO ATTRACT SPECIALISTS WITH AI COMPETENCIES

The need for AI Governance

Al governance is a subset discipline of corporate governance, focused on Al and its performance and ethical risk management



Al adoption and use barriers



The deployment of AI in the public sector requires that several levels are aligned and coordinated, and necessitates several changes at various levels in organizations

Key pillars of governance

Structures

Roles and responsibilities, IT organisation structure, CIO on Board, IT strategy committee, IT steering committee(s)

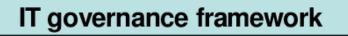
Processes

Strategic Information Systems Planning, (IT) BSC, Information Economics, SLA, COBIT and ITIL, IT alignment / governance maturity models

Al governance can be considered as a sub-set

of IT governance, and can be mapped based

on the three main pillars



Relational mechanisms

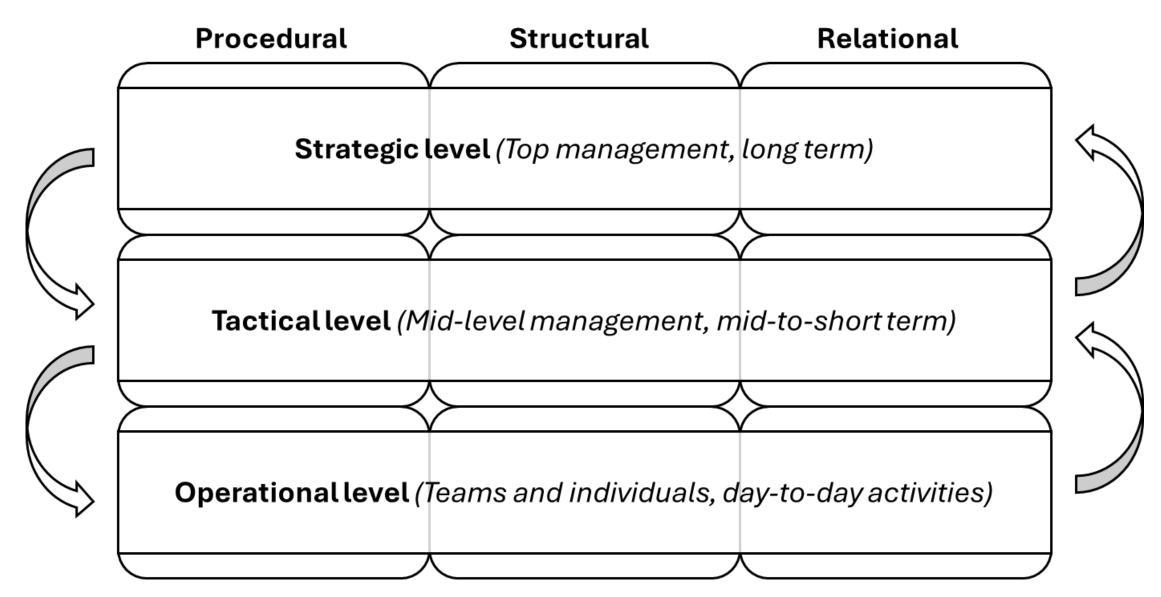
Active participation and collaboration between principle stakeholders, Partnership rewards and incentives, Business/IT co-location, Crossfunctional business/IT training and rotation

De Haes, S., & Van Grembergen, W. (2006, January). Information technology governance best practices in Belgian organisations. In Proceedings of the 39th Annual Hawaii International Conference on System Sciences (HICSS'06) (Vol. 8, pp. 195b-195b). IEEE.

Levels of decision-making

Decision-making around AI takes places at different levels within public organizations, and is typically a responsibility of different actors, with different time-frames associated to each.





Procedural	Structural	Relational	
	\checkmark		
 Developing ethical AI guidelines Compliance protocols Establishing accountability procedures 	 Defining data stewards Establishing independent ethics committees Developing an ethical code of conduct Establishing cybersecurity department 	 Establishing communities of practice Stakeholder education and training Experimentation and idea generation Fostering knowledge transfer 	
\succ	\rightarrow \rightarrow	\rightarrow	
 Minimizing authorization to access data Developing explainability frameworks Monitoring AI usage Developing AI protocols for standardization Ensuring security of algorithmic operation AI lifecycle management processes 	 Safety barriers to prevent misuse Establishing algorithmic registries Defining project ownership Developing steering group Elimination of algorithmic censorship 	 Negotiating and contracting with vendors Promoting society-in-the- loop activities 	
\succ	\rightarrow	\rightarrow	
 Data management Establishing system/and data integration Developing processes for elimination of bias Establishing algorithmic transparency Model reusability 	 Process-based interactions between people and AI End-user participation in AI development and evaluation Ensuring human monitoring and supervision of AI decision-making 	 Promoting collaborative efforts between stakeholders Educating users to develop trust towards AI 	
-	 Developing ethical Al guidelines Compliance protocols Establishing accountability procedures Minimizing authorization to access data Developing explainability frameworks Monitoring Al usage Developing Al protocols for standardization Ensuring security of algorithmic operation Al lifecycle management processes Data management Establishing system/and data integration Developing processes for elimination of bias Establishing algorithmic transparency 	 Developing ethical AI guidelines Compliance protocols Establishing accountability procedures Minimizing authorization to access data Developing explainability frameworks Monitoring AI usage Developing explainability frameworks Monitoring AI usage Developing explainability frameworks Monitoring AI usage Developing al protocols for standardization Ensuring security of algorithmic operation Al lifecycle management Establishing system/and data integration Developing processes for elimination of bias Establishing algorithmic transparency Monitoring algorithmic transparency 	

Key recommendations



Communities of practice Sharing knowledge and

best practices can significantly reduce barriers and uncertainty 02

Provide guidelines

Educational material, seminars, expert consultations, best practices examples



Stakeholders-inthe-loop

Communication channels are necessary to effectively connect with key stakeholders



BREAK SECTION

