

## *Session 5:*

# The OECD AI Systems Classification Framework: progress, challenges & way forward

- Part 1: Overview of the OECD's work on AI
- Part 2: Introducing the classification framework
- Part 3: Using the framework on actual AI systems
- Part 4: Applying the framework to jobs & skills
- Part 5: Discussion



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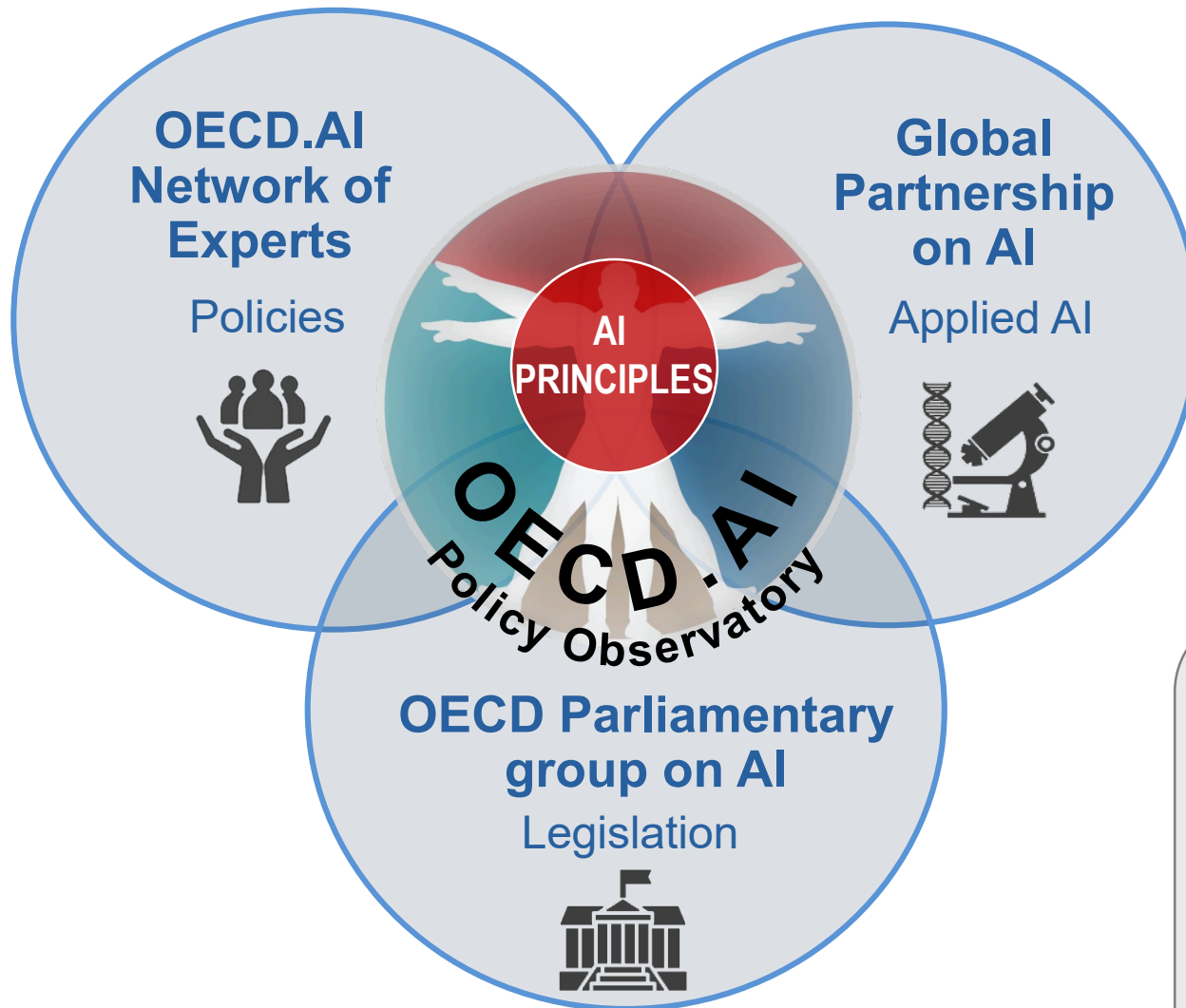
## 10 PRINCIPLES FOR TRUSTWORTHY AI

### Value-based principles

- Socio-economic & environmental impacts
- Human-centred values and fairness
- Transparency, explainability
- Robustness, security, safety
- Accountability

### Recommendations for policy makers

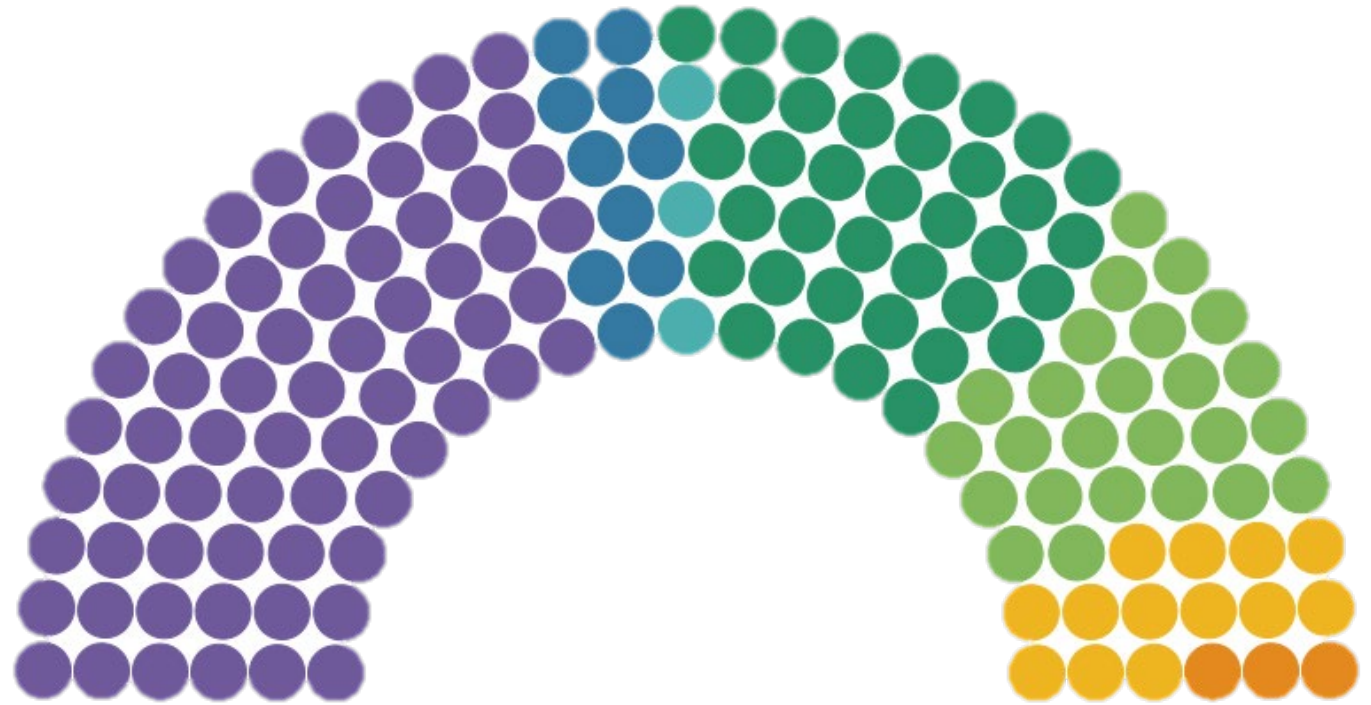
- Investing in research
- Data, compute, technologies
- Enabling policy environment
- Jobs, skills, transitions
- International cooperation



PARTICIPATION  
in complementary  
initiatives by partner  
institutions

- EC
- G20
- UNESCO, UN
- CoE
- IDB, CAF
- ISO, IEEE...

- multi-disciplinary and multi-stakeholder
- **200+ AI experts** from national governments, IGOs and the EC, business, civil society, academia, trade unions.
- advises the OECD & facilitates collaboration with other initiatives.



● Governments ● International Organisations ● European Commission  
● Business ● Civil Society & Academia ● Technical community ● Trade unions

*What types of AI systems raise what types of policy issues?*

**WG CAI :  
Classifying  
AI systems**

## 10 PRINCIPLES FOR TRUSTWORTHY AI

### Values-based principles

- Socio-economic & environmental impacts
- Human-centred values and fairness
- Transparency, explainability
- Robustness, security, safety
- Accountability

### National Policies

- Investing in research
- Data, compute, technologies
- Enabling policy environment
- Jobs, skills, transitions
- International cooperation

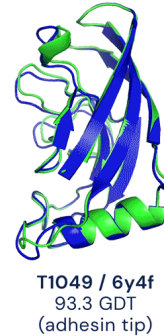
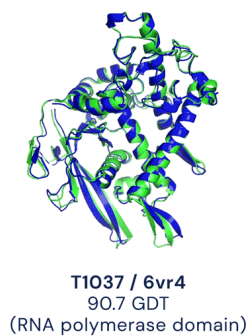
**WG TAI:  
Tools for  
Trustworthy AI**

**WG PAI:  
National  
AI  
policies**

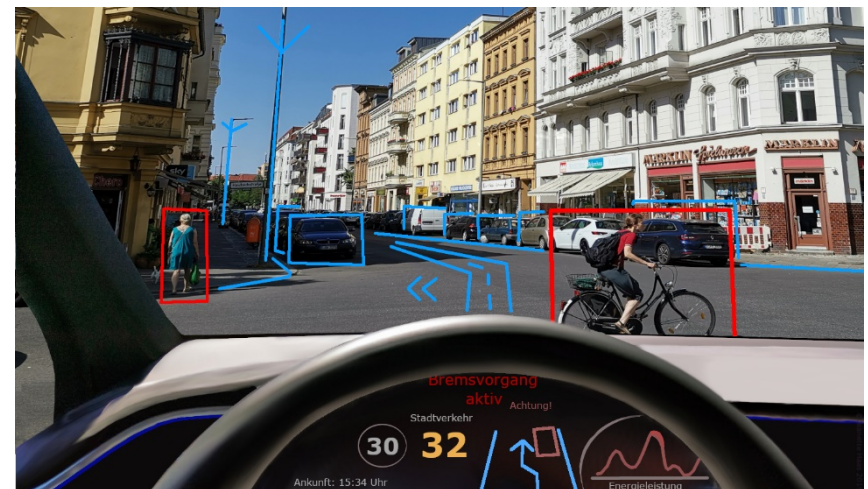
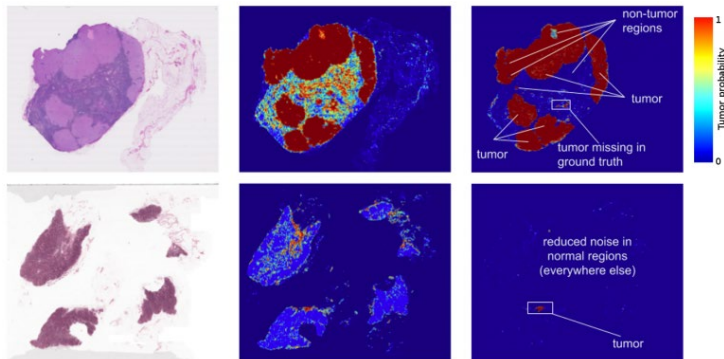
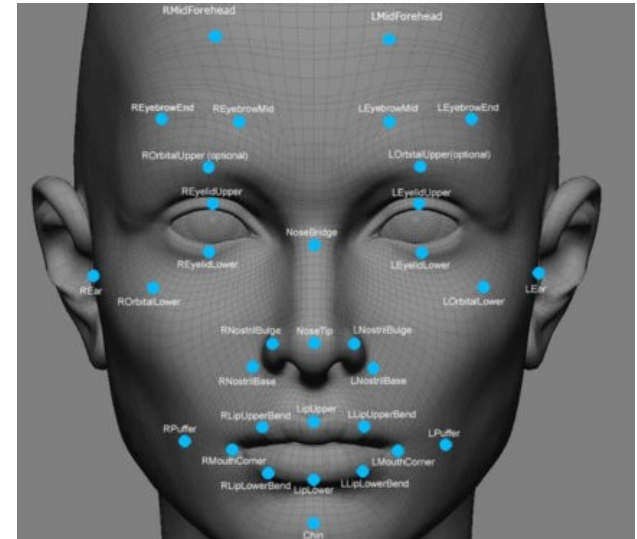
**AI  
Compute**

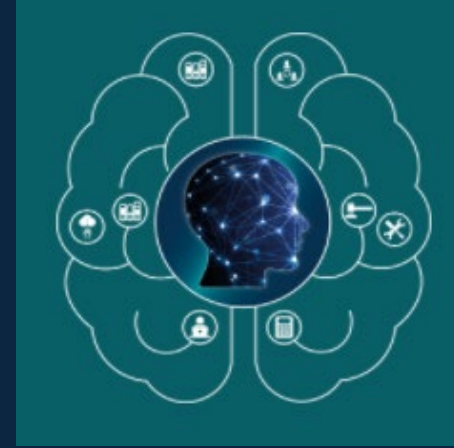


# Why classify AI systems?



● Experimental result  
● Computational prediction





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**Part 2: Introducing the classification framework**

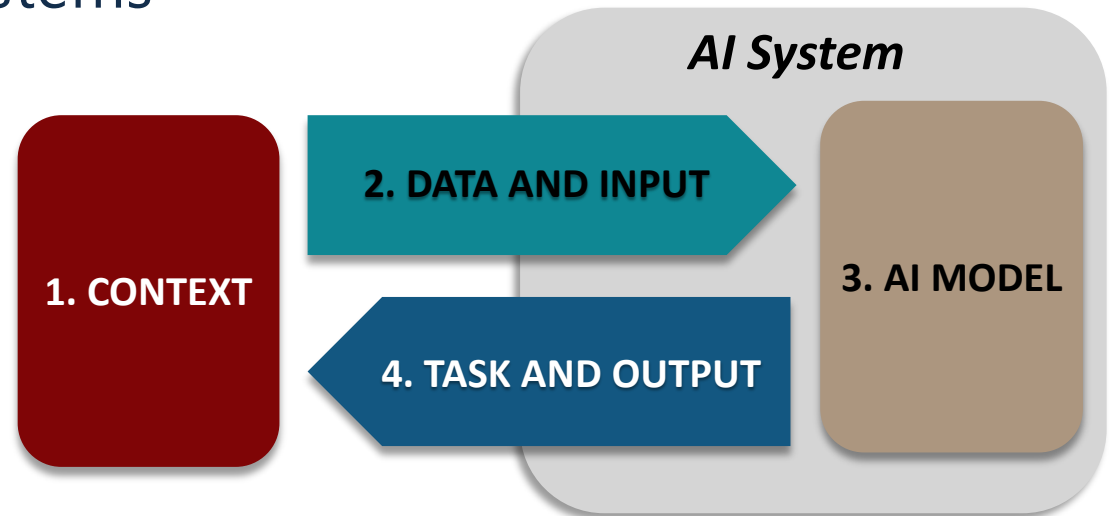
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Part 5: Discussion



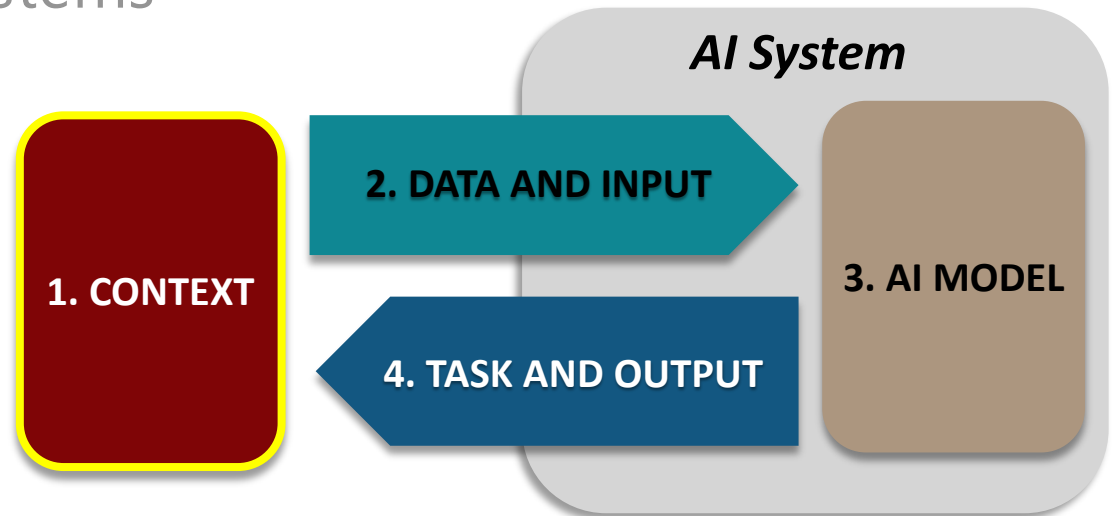
A user-friendly framework to navigate policy implications of different *types* of AI systems



## 4 key dimensions:

1. **Context**, including sector (healthcare, etc.), impact and scale
2. **Data and input**, including data collection, personal nature of data
3. **AI model (technologies)**, incl. model type and model building process
4. **Task and output**, incl. AI system's task (e.g., recognition, personalisation, etc.) and action autonomy

A user-friendly framework to navigate policy implications of different *types* of AI systems

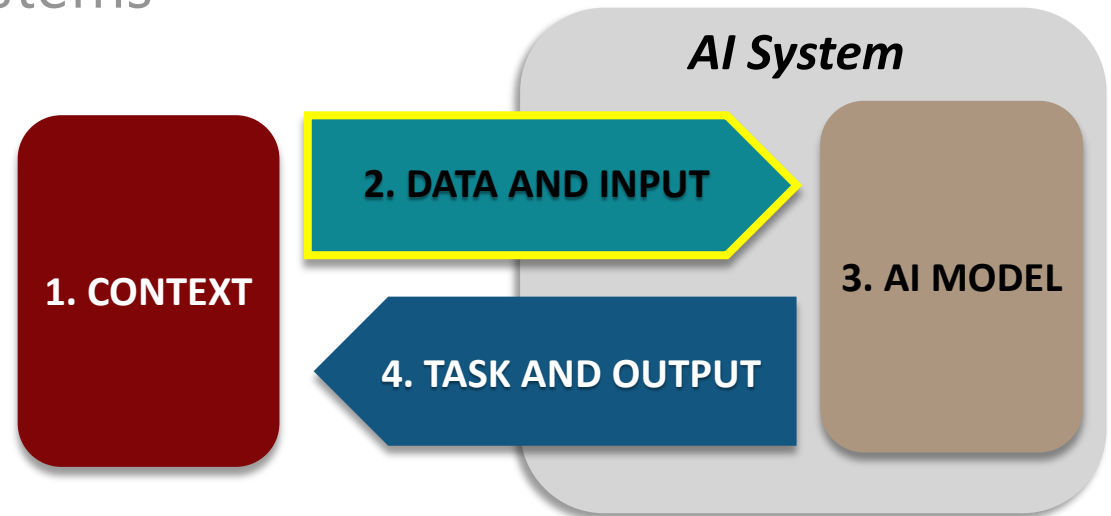


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Properties / Attributes	Description
<b>Industrial sector</b>	<ul style="list-style-type: none"> <li>Using: International Standard Industrial Classification of All Economic Activities (ISIC REV 4) to identify sector and compare to cross-country data</li> </ul>
<b>Deployment impact / scale / impacted stakeholders</b>	<ul style="list-style-type: none"> <li>Breadth of deployment (e.g. pilot, company level)</li> <li>AI system maturity</li> <li>Stakeholders impacted (e.g. workers, consumers)</li> <li>For-profit or non-profit use</li> </ul>
<b>Benefits and risks to individual rights and well-being</b>	<ul style="list-style-type: none"> <li>Impacts on human rights (e.g. criminal sentence)</li> <li>Impact on well-being (e.g. job quality)</li> <li>Low-risk contexts (e.g. restaurant recommendation)</li> </ul>
<b>Benefits and risks to critical functions / activities</b>	<ul style="list-style-type: none"> <li>AI system serving in critical sector or infrastructure (e.g., energy, transport, water, health)</li> <li>AI system serving critical functions in a sector</li> </ul>
<b>Related Consideration</b> (To be developed)	<ul style="list-style-type: none"> <li>Users of AI system (e.g. expert vs. non-expert)</li> </ul>

A user-friendly framework to navigate policy implications of different *types* of AI systems

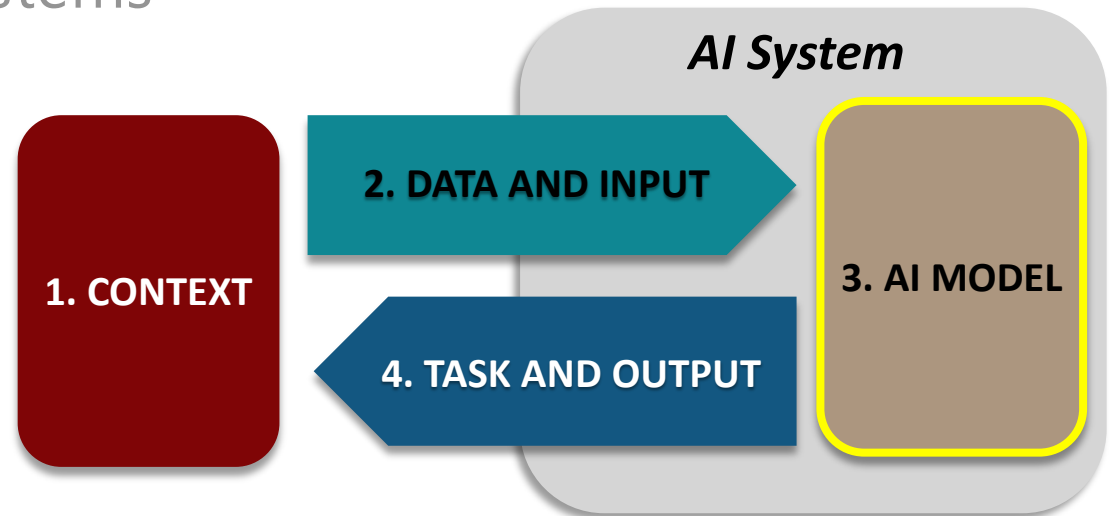


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Properties / Attributes	Description
<b>Data and input collection/origin</b>	<ul style="list-style-type: none"> <li>• Data collection (by sensing or perceiving)</li> <li>• Data provenance (level of awareness)</li> <li>• Dynamic nature of data (e.g. real-time), Scale of data set</li> </ul>
<b>Data and input structure</b>	<ul style="list-style-type: none"> <li>• Structure of data (e.g. structured vs. unstructured data),</li> <li>• Data encoding (standardised format vs. non-standardised format)</li> </ul>
<b>Data and input domains</b>	<ul style="list-style-type: none"> <li>• 3 domains: Proprietary data, Public data, Personal data</li> <li>• Taxonomy of personal data (for assessing the risk to privacy)</li> </ul>
<b>Data quality</b>	<ul style="list-style-type: none"> <li>• Representativeness</li> <li>• Adequate sample rate, Data noise, Missing values and outliers</li> </ul>
<b>Data qualification</b>	<ul style="list-style-type: none"> <li>• Guarantee the appropriate data as input to solve a specific problem (note: this is different from data quality)</li> </ul>
<b>Related Consideration</b> (To be developed)	<ul style="list-style-type: none"> <li>• Implication for privacy and fairness (e.g. sensitive data)</li> <li>• Implication for robustness (e.g. quantity of data vs. the number of variables in a model)</li> </ul>

A user-friendly framework to navigate policy implications of different *types* of AI systems



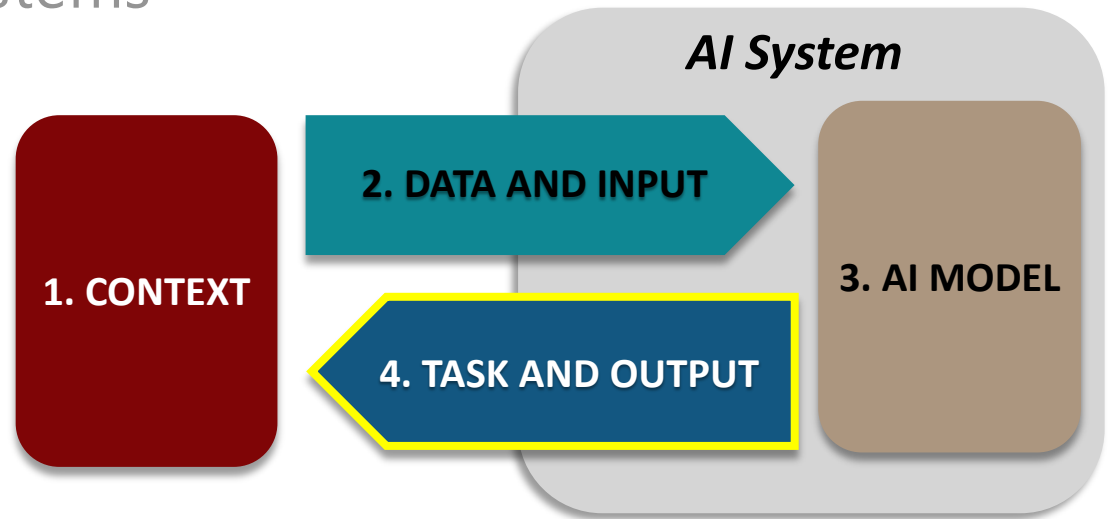
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Properties / Attributes	Description
<b>AI model type</b>	<ul style="list-style-type: none"> <li>• <b>Symbolic AI models</b></li> <li>• <b>Statistical AI models</b></li> <li>• <b>Hybrid AI models (majority of systems live here)</b></li> </ul>
<b>Acquisition of capabilities / model building</b>	<ul style="list-style-type: none"> <li>• Acquisition from knowledge: symbolic AI</li> <li>• Acquisition from data: machine learning</li> </ul>
<b>Related Consideration</b>	<ul style="list-style-type: none"> <li>• <b>Fairness:</b> Choice of AI model, input datasets, and tasks all relate to downstream issues of (potential) bias.</li> <li>• <b>Accountability:</b> This is determined by how humans interact with the model (are they 'on the loop' or 'in the loop').</li> <li>• <b>Robustness:</b> Need for mechanism to ensure the robustness of modern ML systems (e.g. adversarial defenses, fail-safe procedures, formal verification)</li> <li>• <b>Performance:</b> Different models have different performance characteristics</li> </ul>

A user-friendly framework to navigate policy implications of different *types* of AI systems

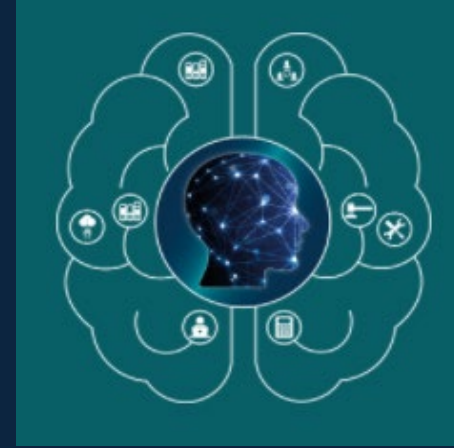


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4. **Task and output**, incl. the AI system's task (e.g., recognition, personalisation, etc.) and action autonomy

# 4) TASK AND OUTPUT

Properties / Attributes	Description
<b>Task of the system</b>	<ul style="list-style-type: none"> <li>• Recognition, Event detection, Forecasting, Personalisation, Interaction support, Goal-driven optimization, Reasoning with knowledge structures</li> </ul>
<b>Action of the system (autonomy level)</b>	<ul style="list-style-type: none"> <li>• High autonomy (<b>human-out-of-the-loop</b>)</li> <li>• Medium autonomy (<b>human-on-the-loop</b>)</li> <li>• Low autonomy (<b>human-in-the-loop</b>)</li> </ul>
<b>Combining tasks and actions into composite systems</b>	<p>Specific policy consideration for:</p> <ul style="list-style-type: none"> <li>• Autonomous systems (e.g. driverless vehicle)</li> <li>• Control systems (e.g. robotics and factories)</li> <li>• Content generation (e.g. misinformation, deep fake)</li> </ul>
<b>Related Consideration (To be developed)</b>	<ul style="list-style-type: none"> <li>• Robustness and security (e.g. expected performance and reliability information)</li> <li>• System operator (e.g. role of evaluator)</li> <li>• Degree of automation (e.g. job displacement)</li> </ul>



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***Disclaimer** : the following slides use existing commercial products for purely theoretical purposes. The OECD classification framework has not coordinated with these products' owners. As a result, there may be inaccuracies based on the authors' read of the publicly accessible materials.*

1

## Sector of deployment

(e.g., Transportation and storage, human health and social work activities, Education)

## Critical function

(e.g., health, safety, and security of citizens; essential economic and societal services)

## System users

(e.g., AI-expert vs. non-AI expert user)

2

## Data collection

(e.g., humans, automated, system experience)

## Data domain

(e.g., proprietary, public, personal)

## Data structure

(e.g. structured, semi-structured, unstructured)

3

## Acquisition of capabilities

(e.g., learn from people vs. provided data vs. system experience)

## AI model type

(e.g. supervised, probabilistic, symbolic)

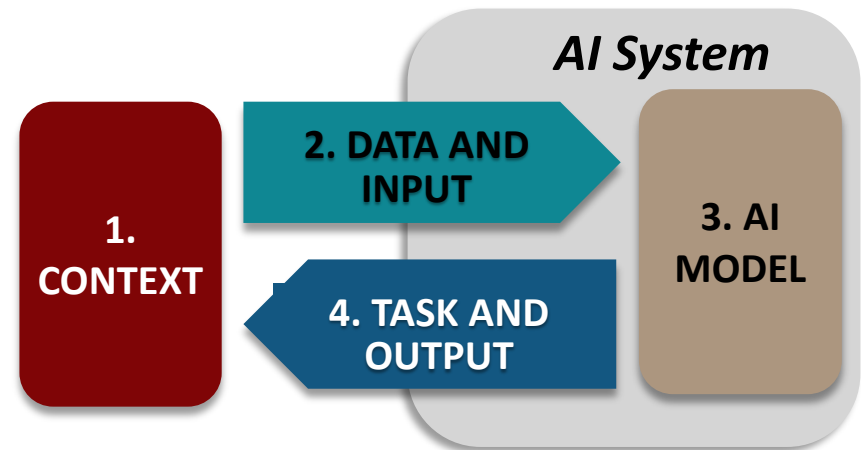
4

## System task

(e.g., recognition, personalization, goal-driven optimization)

## Level of action autonomy

(e.g., high (human out-of-the-loop), medium (human on-the-loop), low (human-in-the-loop))







**Development practitioners need a decision-making framework to determine whether systems can effectively support development objectives in order to invest.**

This framework, **when used responsibly**, can help in

- Mapping **value propositions of AI systems to development outputs.**
- Reflect on potential **challenges and risks** ( in design, implementation, impacts)
- Improving regulation, tech policy and **impact.**

## Sector of deployment

Education

1

## Critical function

No

## System users

Non AI expert amateur (students)

## Data collection

Collected by humans and automated tools

2

## Data domain

Personal

## Data structure

Structured

## Acquisition of capabilities

From data

3

## AI model type

Generative, semi-supervised, non-probabilistic

4

## System task

Personalization, interaction support

## Level of action autonomy

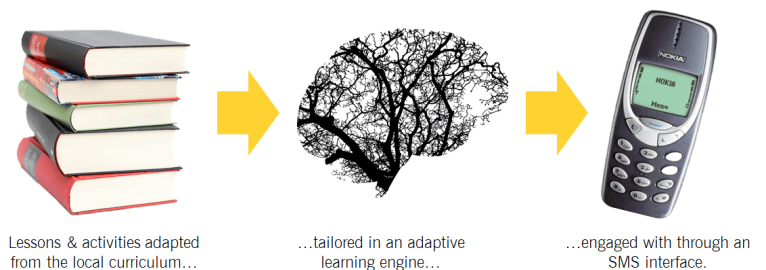
Medium – provides recommendation for human decision execution

### Example ID

**Name:** M-Shule

**Link:** [here](#)

**Short description:** Delivers personalized lessons via SMS based on curriculum standards and student skills and abilities. Tracks and analyses performance to provide recommendations.



1

## Sector of deployment

Administrative and support service activities

## Critical function

No

## System users

Amateur (job candidate and HR personnel)

2

## Data collection

Collected by automated tools

## Data domain

Proprietary, proprietary personal

## Data structure

Unstructured

3

## Acquisition of capabilities

From data

## AI model type

Discriminative, semi-supervised, probabilistic

4

## System task

Personalization, interaction support, recognition

## Level of action autonomy

Medium – provides recommendation for human decision execution

### Example ID

**Name:** myInterview

**Link:** [here](#)

**Short description:** conducts smart video interviews, picking questions based on company needs, and scans all interview videos to create curated interview “playlists” for companies to review.

Framework  
Dimensions:

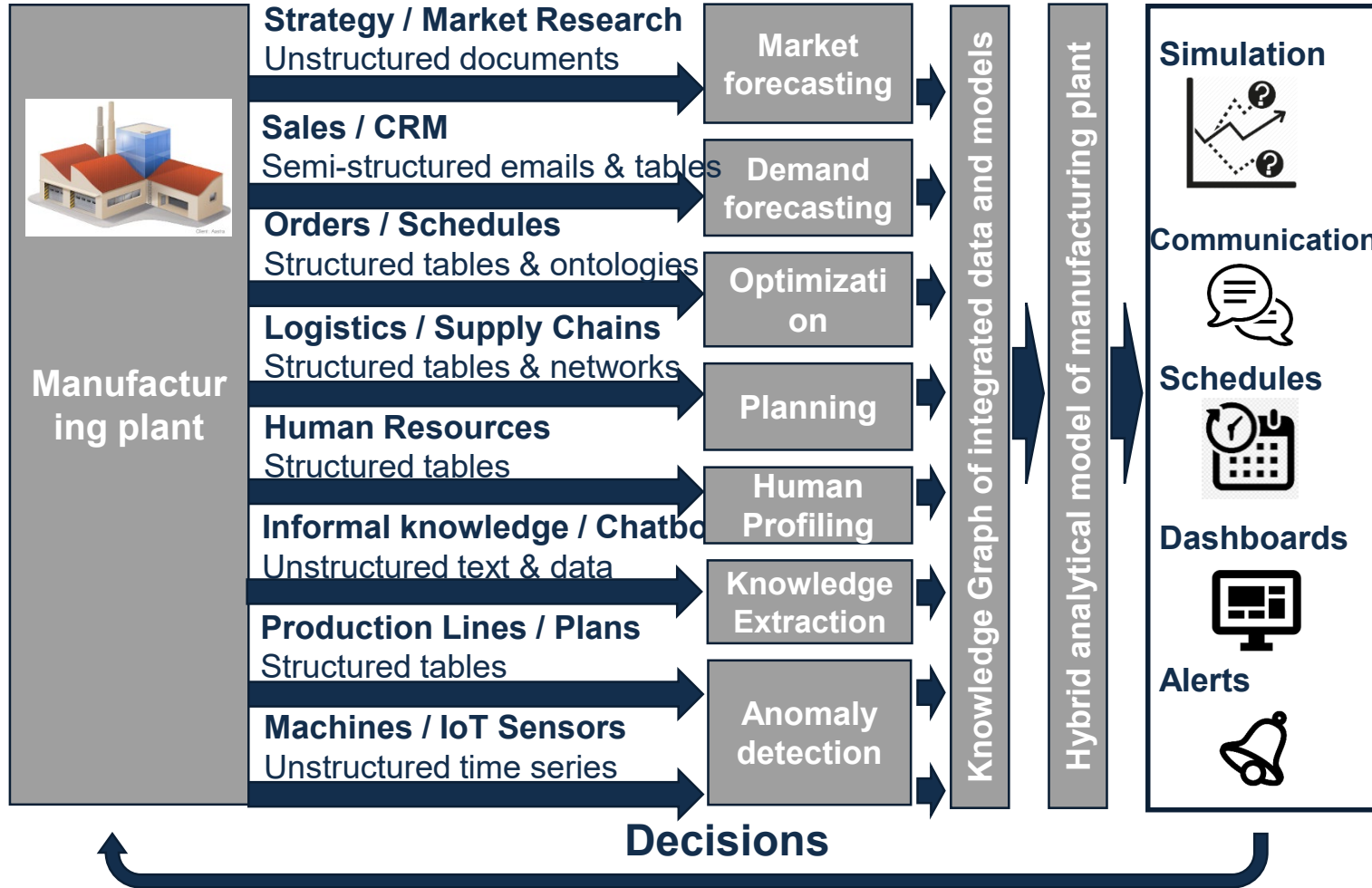
**1. Context**

**2. Data & Input**

**3. AI Model**

**4. Output**

**Manufacturing  
plant  
example  
structured  
using OECD  
AI Classification  
Framework**



1

## Sector of deployment

Public administration and defense

## Critical function

Yes

## System users

Practitioner who is not an AI expert (police)

## Data collection

Collected by automated tools

2

## Data domain

Public

## Data structure

Semi-structured

3

## Acquisition of capabilities

From data augmented by human-encoded knowledge

## AI model type

Hybrid model

4

## System task

Recognition

## Level of action autonomy

Medium - provides a prediction for human decision execution.

### Example ID

**Name:** ShotSpotter

**Link:** [here](#)

**Short description:** ShotSpotter is a technology deployed in 85 cities which uses a combination of acoustic microphones and software to automatically listen for gunfire and, if it thinks it has heard gunfire.

## Sector of deployment

Public administration and defense

1

## Critical function

Yes

## System users

Practitioner who is not an AI expert (justice professionals)

## Data collection

Collected from human inputs

2

## Data domain

Personal

## Data structure

Structured

3

## Acquisition of capabilities

From data and human-encoded knowledge

## AI model type

Generative, supervised learning, non-probabilistic

4

## System task

Forecasting, recognition

## Level of action autonomy

Medium - provides a prediction for human decision execution.

### Example ID

**Name:** ML algorithms for predicting juvenile recidivism

**Link:** [here](#)

**Short description:** Tool that generates risk assessments based on information on defendant demographics and criminal history.





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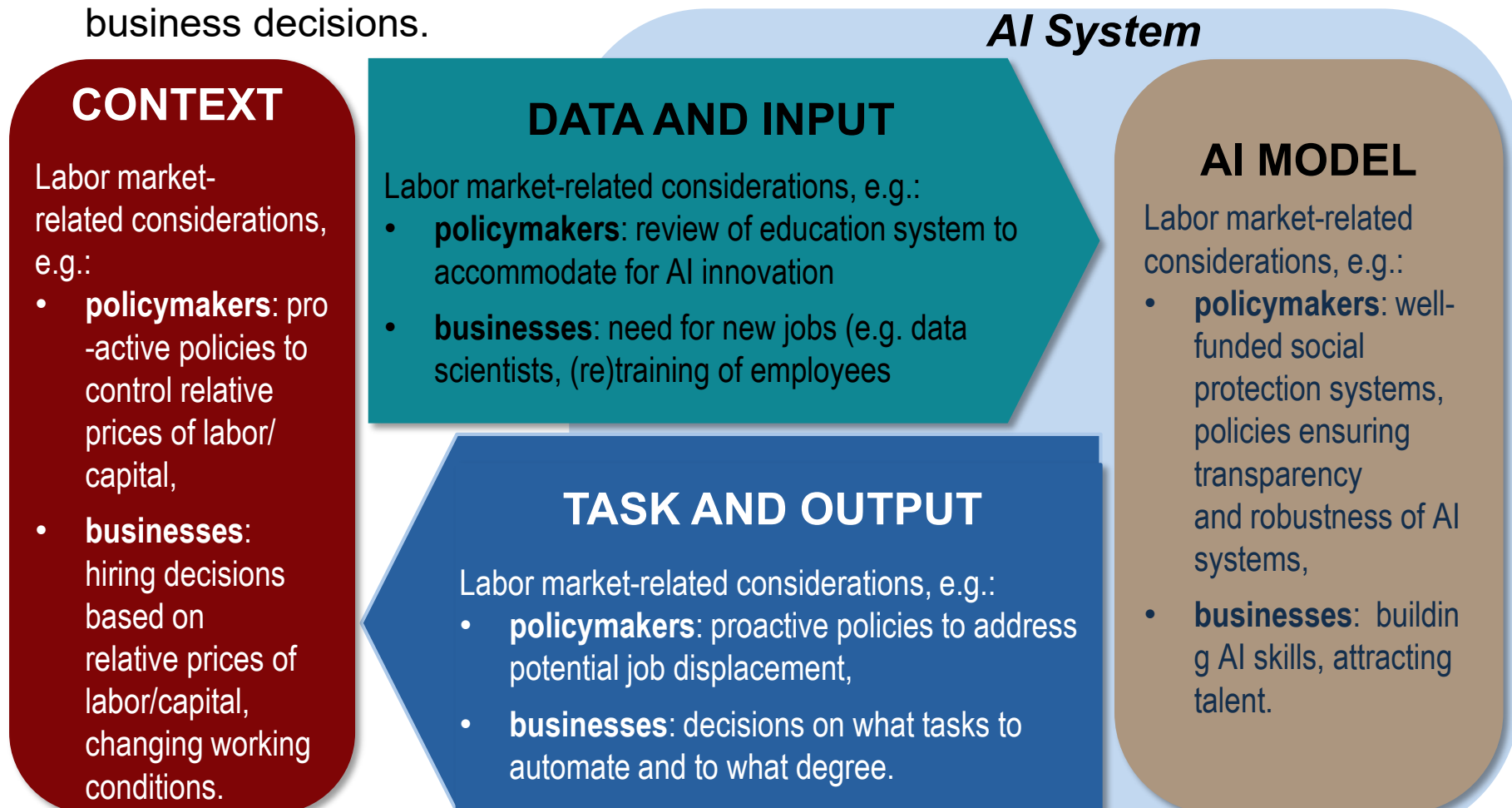
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# Applying the framework to assess labour market impacts

- **Using the framework as a single reference point for all stakeholders**, policymakers: policy considerations, businesses: AI-related business decisions.

- Using the framework as a single reference point for all stakeholders, policymakers: policy considerations, businesses: AI-related business decisions.



# Applying the framework to assess labour market impacts

- **Key issue:** Controlling technological innovation's impact in imperfect markets

## ➤ **Key issue:** Controlling technological innovation's impact in imperfect markets

- **AI is the latest wave of innovation so history can help:**
  - as a general-purpose technology, AI threatens (aspects of) jobs in all segments of economy,
  - it is difficult to anticipate impacts due to **differences to previous waves of innovation.**

## ➤ **Key issue:** Controlling technological innovation's impact in imperfect markets

- **AI is the latest wave of innovation so history can help:**
  - as a general-purpose technology, AI threatens (aspects of) jobs in all segments of economy,
  - It is difficult to anticipate impacts due to **differences** to previous waves of innovation.
- **Much depends on how we control AI innovation:**
  - in perfectly competitive markets innovation would make everyone better off (Pareto-improving).
  - but markets imperfect and structured by regulation,
  - market-structuring, regulations and market imperfections crucially determine the benefit society derives from innovation
  - Pareto-improving innovation is only possible with **redistributive regulation** (otherwise inequality rises/welfare decreases and AI will only amplify those effects).

➤ **Key issue:** Controlling technological innovation's impact in imperfect markets

- **Economy and market-structuring regulation form an interconnected system:**
  - ex post redistribution/piecemeal regulatory approaches merely symptomatic/futile treatment,
  - may even upset economic balance.

## ➤ **Key issue:** Controlling technological innovation's impact in imperfect markets

- **Economy and market-structuring regulation form an interconnected system:**
  - ex post redistribution/piecemeal regulatory approaches merely symptomatic/futile treatment,
  - may even upset economic balance.
- **A truly efficient solution only by systematic revamp of entire system of rules structuring economy:**
  - tax policies, competition laws, IPR, financial regulation,
  - corporate governance, international trade, monetary policy,
  - labor market-related rules, structural discrimination,
  - globalization, innovation, demographics, climate change...



➤ **Key issue:** Controlling technological innovation's impact in imperfect markets

- **Globalization poses extra challenge:** only **internationally coordinated** regulatory efforts efficient (domestic approaches may create inefficiencies/conflicts).

➤ **Key issue:** Controlling technological innovation's impact in imperfect markets

- Globalization poses extra challenge: only internationally coordinated regulatory efforts efficient (domestic approaches may create inefficiencies/conflicts).
- **AI classification framework** could serve as **common benchmark** helping international/multi-stakeholder **coordination**.

## What?

A **framework** for classifying AI according to labour market impact

## Why?

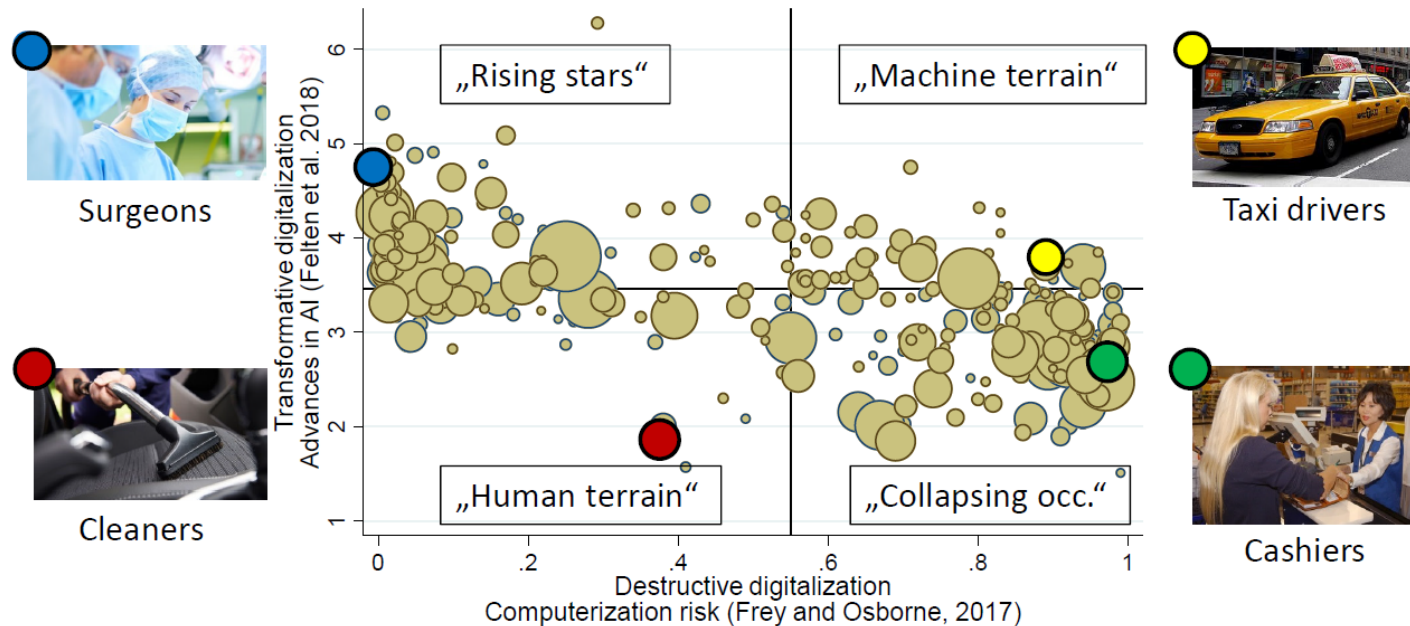
To delve deeper into the likely impact of AI on **labour market outcomes**: wages, employment, the work environment, inclusiveness...

## For whom?

Policymakers, businesses, unions, workers, researchers...

## By whom?

Researchers in the OECD Directorate for Employment, Labour and Social Affairs



Source: based on Fossen and Sorgner (2019)

## How is it related to the AI Systems Classification Framework?

- Matching AI System's tasks with job-specific tasks (e.g. image recognition, event prediction)
- AI System's level of autonomy may create destructive effects (depends on the relevance of the task in a job).
- Effects of AI on labour markets may depend on the context (e.g., industry, region, formal institutions)



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Visit the **OECD AI Policy Observatory** ([OECD.AI](#)) for more, including:

- The OECD programme on Work, Innovation, Productivity & Skills: [oecd.ai/wips](#)
- The AI Wonk blog: [oecd.ai/wonk](#)
- Network of Experts: [oecd.ai/network-of-experts](#)
- Trends and data: [oecd.ai/trends-and-data](#)
- Latest research in AI policy: [oecd.ai/policy-areas](#)
- Database of country AI initiatives: [oecd.ai/dashboards](#)