



Digital Workshop

Artificial Intelligence in Risk Analysis: Balancing Automation and Human Judgement

Group Discussion

Challenges and Lessons Learned for Efficient Deployment of AI Solutions in Risk Analysis



Feedback from Group 1

Is AI currently used in your tax administration for risk analysis?



Current use of AI in tax administrations for risk analysis

› Support functions first

Ireland: not yet using AI for risk analysis, but already using AI for call routing, internal assistance and call summaries.

› Risk analysis is cautious

Denmark: uses risk-related models in complex fraud work, but machine learning is limited because training data are incomplete and fraud patterns evolve quickly.

› Common direction of travel

Other administrations described pilots or early-stage projects, including machine-learning models based on historical audit data and AI-supported information searching.

› Big picture

Italy was described as having a strong data-science capacity, while Latvia and others are still building experience in practical risk-analysis use cases.



Feedback from Group 1

What standards should apply to data quality and reliability?

Standards for Data Quality When Using AI

- › Predefined thresholds for matching accuracy and consistency
- › Documented provenance and validation of external data sources
- › Checks that data granularity and timing fit the analytical purpose
- › A rule that uncertain or unfit sources should not enter the model



Validation & Human Review

Feedback from Group 1

How should auditors validate AI-generated risk signals?

Should AI-generated alerts always be reviewed by a human before action is taken?

➤ Practical rule

Validation should be continuous, expert-led and closely tied to real operational cases.

➤ Recommended review flow

- ✓ Model produces a signal
- ✓ Auditor receives context or explanation
- ✓ Signal is checked against real case evidence
- ✓ Action follows only after human assessment



Feedback from Group 1

How should tax administrations detect and mitigate algorithmic bias in risk models?

Detecting and Mitigating Bias in Risk Models

➤ Beyond Historical Data

Training only on known historical cases can reproduce existing detection patterns and miss new fraud types.

➤ Bias Mitigation Approach

Bias mitigation, therefore, requires careful review of training data, ongoing monitoring and strong expert challenge.

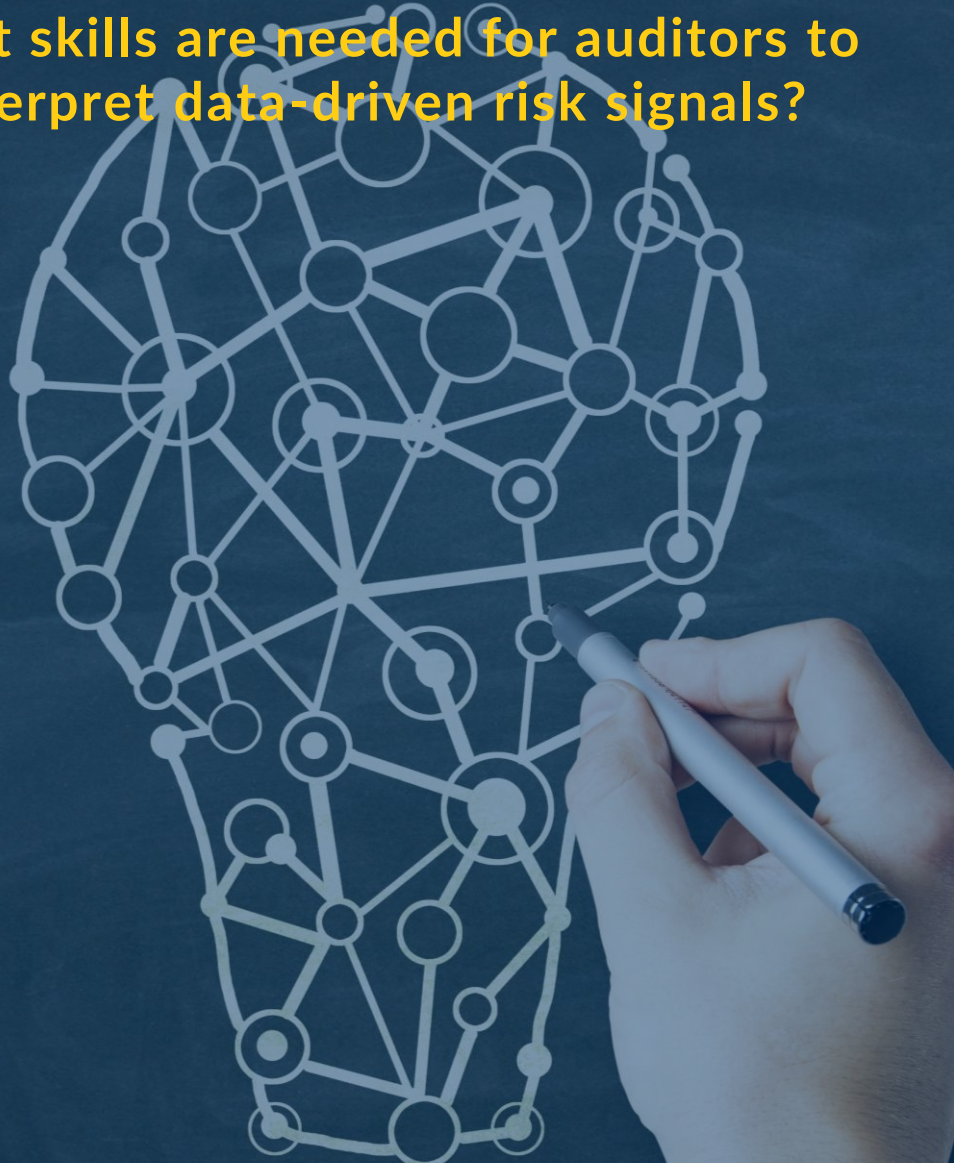
➤ Keep an eye on divergences

Looking closely at cases where AI and human judgment diverge, and building models that provide some explanation of why a case was flagged.



Feedback from Group 1

What skills are needed for auditors to interpret data-driven risk signals?



Skills auditors need

- › Strong tax and fraud expertise, plus the ability to interpret risk signals critically
- › Understanding of data limitations and enough analytical literacy to ask the right questions
- › Capacity to work with analysts and technical teams in multidisciplinary settings
- › Use of tools such as crime scripts to translate real fraud patterns into model logic



Feedback from Group 1

How much transparency is appropriate without compromising fraud detection methods?

Human Oversight, Taxpayer Rights & Transparency

› Human in the loop

AI may prioritise or triage, but major compliance action should remain human-led.

› Explain actual cases

Formal case decisions need proper human explanation and documentation.

› Do not reveal the playbook

Avoid exposing detailed model logic that could become a manual for fraudsters.



Feedback from Group 1

What governance structure should exist?

Governance Structures for Responsible AI Use

- **Legal Basis**
AI use should rest on a clear legal framework, with the law updated where necessary.
- **Privacy Safeguards**
Use anonymised data where possible and protect sensitive information throughout development and evaluation.
- **Independent Oversight**
Parliamentary auditors or equivalent public bodies should provide external assurance.
- **Human Accountability**
Consequential action must remain attributable to human officials.
- **Model Validation Function**
A dedicated validation team or peer-review mechanism should challenge technical robustness before and after deployment.
- **Real-World Evaluation**
Meaningful evaluation should be based on real operational data, where lawful and possible, ideally anonymised.
- **Cross-Functional Dialogue**
Leadership, legal teams and analysts need a shared understanding of model-building choices.



The Path Forward

- › **Start with reliable data**
Set entry standards for sources and exclude what is unfit.
- › **Build explainable signals**
Use modular, expert-informed models that can be validated.
- › **Keep humans accountable**
Reserve consequential decisions for trained officials.
- › **Scale under governance**
Protect privacy, monitor bias and embed independent scrutiny.

