

Context

Understanding the urgency of AI governance in taxation

AI is entering core tax workflows

- Risk assessment & stratification
- Anomaly detection
- Virtual assistants
- Document drafting
- Data extraction & classification

Regulation is catching up

- OECD AI Accountability guidance
- EU AI Act
- Expanding digital reporting regimes

Rising expectations from authorities & boards

- “We should get instant analysis in today’s world”
- “Show me your workings.”
- “Explain the AI-assisted position.”

Key Insight: Tax authorities worldwide are already using AI for fraud detection and risk scoring. Tax teams must ensure their AI outputs are **defensible, not just efficient**.

The Tax Trust Gap



AI is Powerful - But Opaque

Modern AI systems, especially large language models, operate as "black boxes" making decisions that even their creators cannot fully explain and concerns around safety & security.



Tax Demands Transparency

- Repeatability of calculations
- Defensibility before authorities
- Complete traceability of decisions

Professionals will now need to interrogate the system to understand why a model arrived at a conclusion.

The Core Question

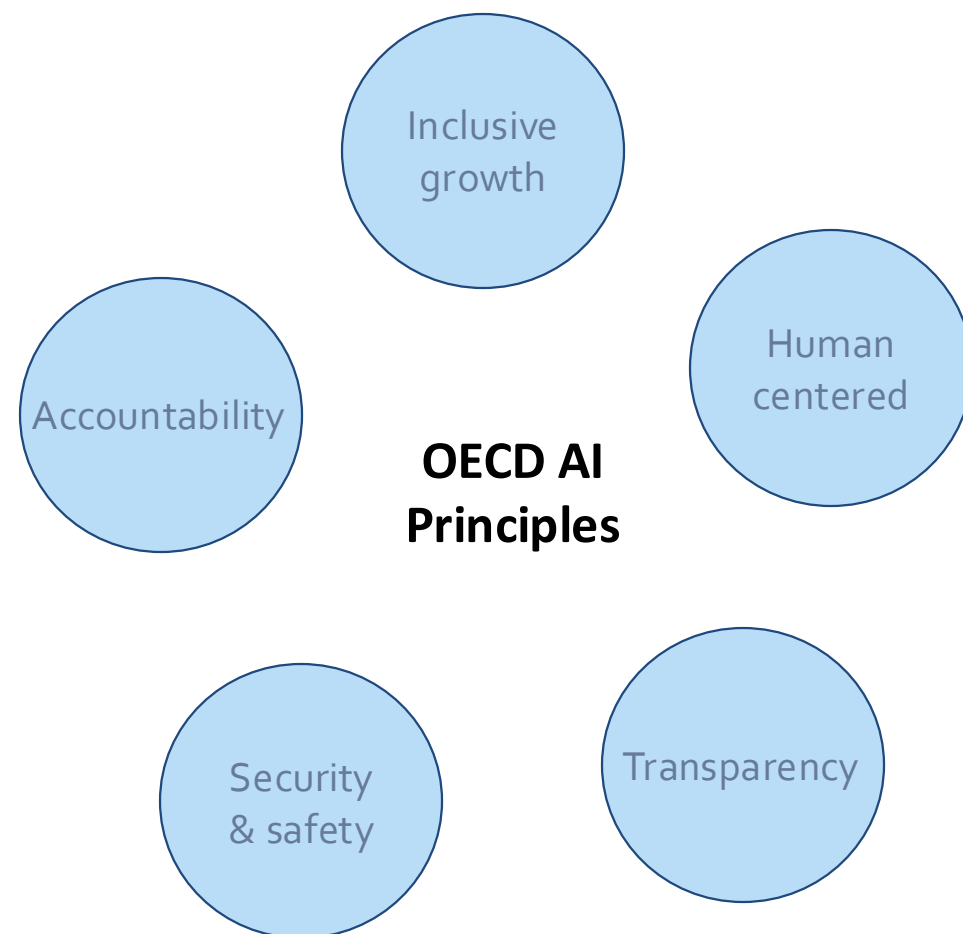
Can you show **how** you arrived at the conclusion?

OECD Alignment: Direction of Travel

How international standards shape AI governance in tax

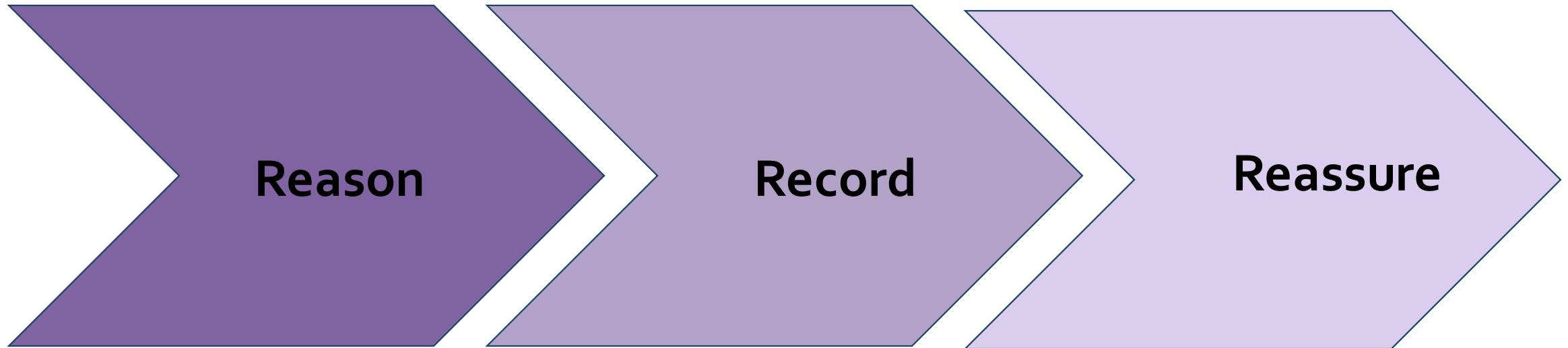
OECD AI Principles (2024)

The OECD's updated AI principles emphasize that high-impact domains require explainable and reconstructible AI systems.



The 3R Accountability Framework for AI in Tax

A practical model for responsible AI in tax



This framework lets tax functions adopt AI without losing control

Pillar 1: Reason

Key Requirements

1. Clear Rulebook

Define explicit boundaries for interpretation

2. Approved Sources

Specify authoritative data and legal references

3. Policy Alignment

Ensure AI follows organizational tax policies

4. Human Escalation

Route grey areas to professionals

Why Reason Matters

AI can handle volume but struggles with nuance.

Defining rules, thresholds, and allowed sources ensures the system's behavior matches tax logic, not creative inference.

Human Judgement needed

When AI encounters ambiguity, uncertainty, or novel scenarios, it must flag for human review rather than proceeding autonomously.

Reason: Standardizing the workplace

What must be standardised is the AI-enabled workspace, not the intelligence itself

Standardize

Input formats & templates

Mapping rules & logic

Adjustment lists

Validation criteria

Allow Controlled Flexibility

Reviewer overrides

Interpretation notes

Exception routing

Commentary fields

Pillar 2: Record

Every action the agent takes is logged: which data sources it accessed, which analysis framework it applied

Prompts & Inputs

All queries sent to AI systems

Model Versions

Specific versions used for each task

Data Lineage

Source and transformation of data

Applications

Which rules triggered and why

Overrides

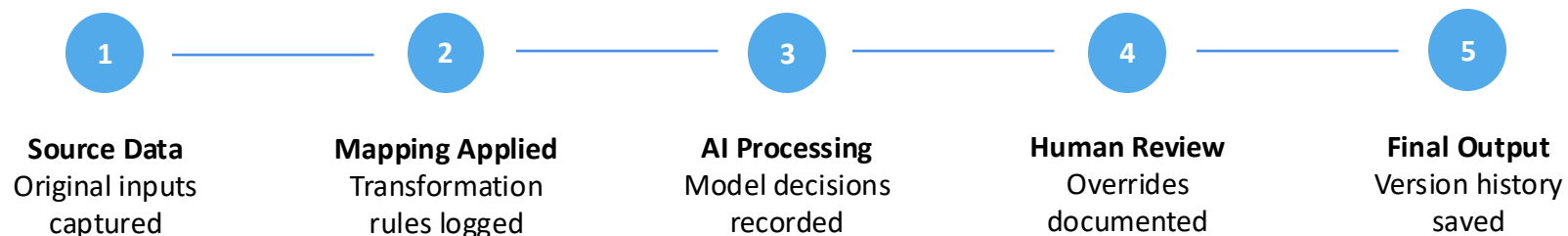
Manual adjustments with rationale

Why Record Matters

Without logs, you cannot defend positions. Without traceability, auditors will not accept AI-enabled steps. Record is where **accountability becomes auditable**.

Record: What a Good Audit Trail Looks Like

Creating defensible documentation



Component	What to Capture	Purpose
Source Data	Raw files, timestamps, checksums	Prove data integrity
Rule Applications	Which rules fired, parameters used	Explain calculations
Reviewer Notes	Comments, rationale, concerns	Document professional judgement
Evidence Pack	Supporting documents, references	Substantiate positions
Version History	All changes with timestamps	Enable reconstruction

Pillar 3: Reassure

Make system behavior understandable to all stakeholders



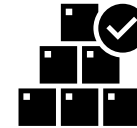
Interpretability
Plain-language
explanations of AI
decisions



**Confidence
Statements**
Probability scores
and uncertainty
flags



Role Clarity
Clear ownership
and accountability
chains

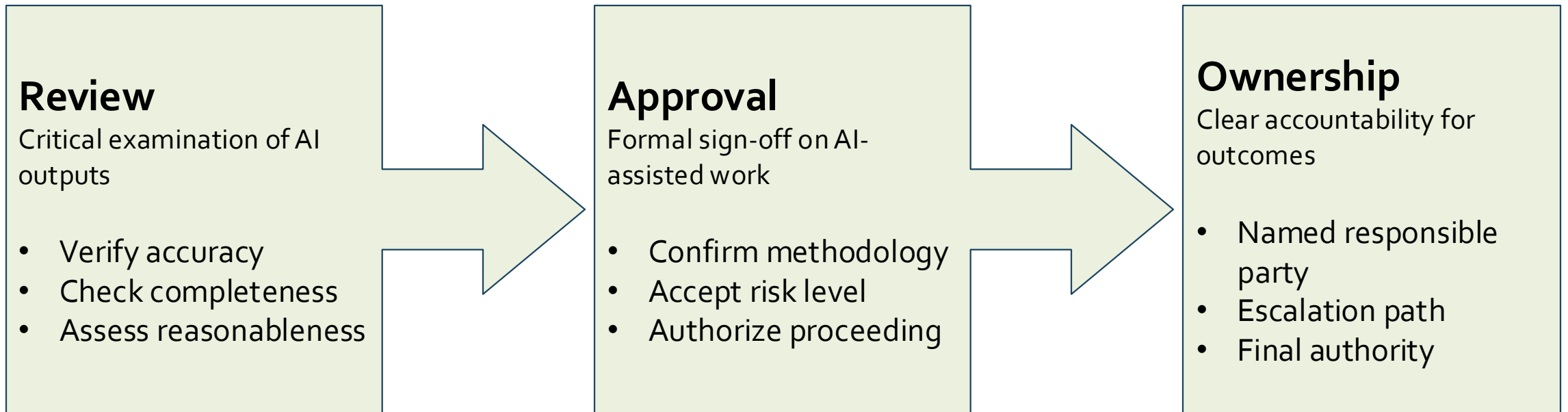


**Decision
Explanations**
Step-by-step
reasoning
documentation



Stakeholder Docs
Board and
auditor-ready
reports

Accountability Nodes



Individuals responsible for reviewing, signing off, and taking ownership of outputs form the backbone of AI accountability.

A Unified AI-Enabled Tax Workflow



AI Preparer
Automated data
processing



AI Reviewer
Automated validation



Human Reviewer
Professional oversight



Sign-off
Final approval

Fully Traceable
Every step documented

Repeatable
Consistent results

Defensible
Audit-ready output

AI Use Cases Across the Tax Function



Corporate Tax Preparation
Automated computation and adjustments



GST/VAT Classification
Smart categorization of transactions



WHT Automation
Withholding tax rate determination



Transfer Pricing Schedules
TP documentation and analysis



Notice Response Drafting
AI-assisted reply preparation



Risk Scoring
Anomaly detection & risk stratification



Tax Control Framework
Continuous monitoring & compliance

Deep Dive: Corporate Income Tax

Why CIT preparation is ideal for AI automation

CIT Preparation

- Rule-heavy:** Clear legislative requirements
- Highly structured:** Standard forms and schedules
- High volume:** Many entities, many periods
- Year-on-year consistency:** Similar patterns repeat
- Strong audit requirements:** Documentation essential

OECD Perspective

The OECD identifies data preparation as one of the most automatable tasks in tax administration.

However, every extraction, mapping, and flag must be logged, otherwise this introduces blind spots.

CIT Deep Dive: Data Gathering

STEP 1

How AI handles data preparation



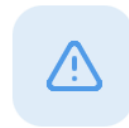
TB / GL Extraction
Automated data pull
from accounting
systems



Consistency Checks
Validation against
prior period data



Mapping to Schedules
Intelligent
categorization



Missing Data Flags
Automated gap
identification



Prior Year Comparison
Variance analysis and
trending

AI Preparer Actions

- ✓ Extract trial balance from ERP
- ✓ Map GL accounts to tax schedules
- ✓ Flag incomplete or unusual entries
- ✓ Compare with prior year patterns

Record Requirements

Every extraction, mapping, and flag must be logged with timestamps, source references, and the specific rules applied. This creates the foundation for an auditable workflow.

CIT Deep Dive: Reason Layer

STEP 2

The system applies approved tax rules across key computation areas:

Depreciation

- Asset categorization
- Rate application
- WDV calculations

Disallowances

- Personal expenses

Interest Limits

- Thin cap rules
- ALP requirements

Loss Set-offs

- Carry forward
- Set-off sequence
- Time limits

Timing Adjustments

- TDS compliance
- Payment dates

Prior-Year Consistency

The system validates that rule applications are consistent with prior year treatments, flagging any changes for human review and documentation of the rationale

CIT Deep Dive: Record Layer

STEP 3

Building the audit-ready evidence pack

Log Category	Data Captured	Audit Purpose
Inputs	Source files, timestamps, checksums	Prove data integrity
Transformations	Mapping rules, calculation steps	Explain methodology
Model Version	AI model ID, parameters used	Enable reconstruction
Rule Applications	Sections applied; thresholds checked	Justify positions
Overrides	Manual changes with rationale	Document judgement
Evidence Pack	Supporting documents, references	Substantiate claims
Reviewer Notes	Comments, concerns, approvals	Record oversight

CIT Deep Dive: Human Reviewer Node

STEP 4

Where professional judgement takes over



Compare with FS

Reconcile with financial statements



Review Anomalies

Investigate flagged items



Validate Rules

Confirm correct rule application



Add Notes

Document interpretations



Approve/Override

Accept or modify AI output



Final Sign-off

Take accountability

The reviewer can cross-check with prior filings, detect red flags, or reconcile with financial statements

Minimum Viable Controls (MVC)

Scaling controls with AI maturity



Maturity →  Sophistication

The AI-Ready Tax Professional

Tax knowledge alone is no longer enough



Deep Tax Knowledge

Core foundation remains essential



Process Understanding

Know how workflows operate



Data Literacy

Understand data structures & quality



Interpreting AI

Evaluate algorithmic outputs



Challenging AI

Question automated decisions

Model Literacy

The future professional doesn't need to code.

They need the ability to:

- Question algorithmic choices
- Understand model limitations
- Identify when AI may be wrong
- Document AI-assisted reasoning

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