

IOE



Investing in rural people

Independent Office of Evaluation

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AI

WEBINAR-TRAINING

Online,
22 September
2025

AI IN EVALUATION:

Lessons from IFAD IOE and Practical Skills for Evaluators

SPEAKERS:

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Senior Evaluation Officer at IFAD's
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Evaluation and AI Specialist at IFAD's
Independent Office of Evaluation (IOE)



MODERATOR: **Lea Corsetti**

CGIAR IAES Consultant; EvalYouth Vice-Chair
(2025-2027) and co-lead of the European Evaluation
Society's Young and Emerging Evaluators group



HOST: **Innocent Chamisa**

EvalforEarth Coordinator





The 'Why': Why AI? Why Now for Evaluation?



The 'What': IOE Use Cases (Successes & Failures)



The 'How': Building the Skills



The 'So What': Lessons Learned, Navigating Risks and Ethics

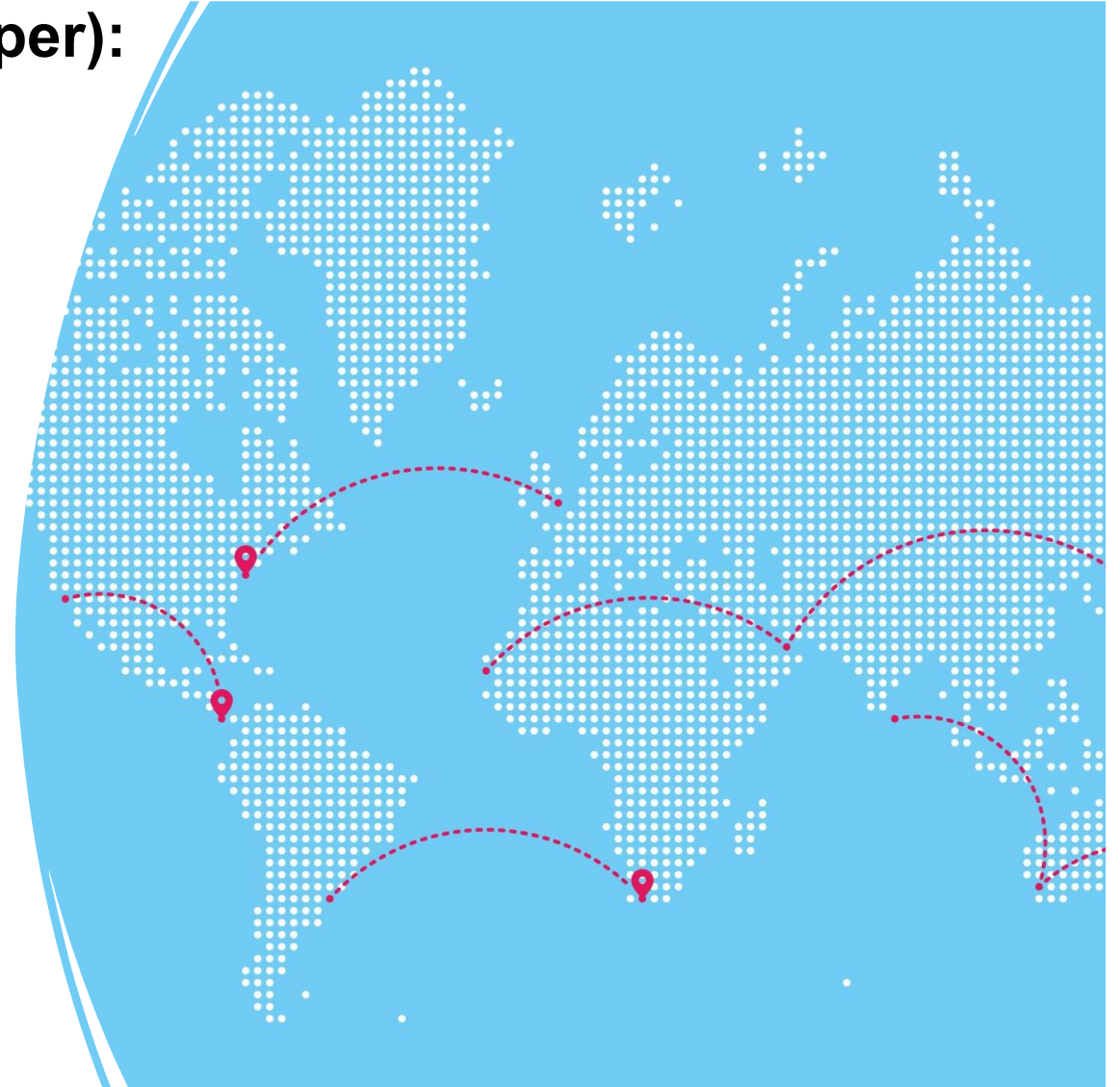
Key Global Statistics (From the NBER Paper):

https://www.nber.org/system/files/working_papers/w34255

- **700 Million+ Users:** ~10% of the global adult population uses OpenAI models weekly.

The Top Tasks

- **#1 Work Use Case: Writing (40% of work messages)**
 - Emails, reports, communications, editing, summarizing.
- **#1 Overall Use Case: Practical Guidance & Decision Support**
 - Asking for advice, how-to information, and customized plans.



The Evaluation Challenge and our commitment: More Data, Less Time, Deeper Questions



An Expanding Evidence Universe

Growing volumes of unstructured data from interviews, reports, projects, synthesize.



Demand for Efficiency & Traceability

Demand to deliver robust insights faster and with clearer audit trails of the analytical process.



Support Evaluators in Repetitive Tasks

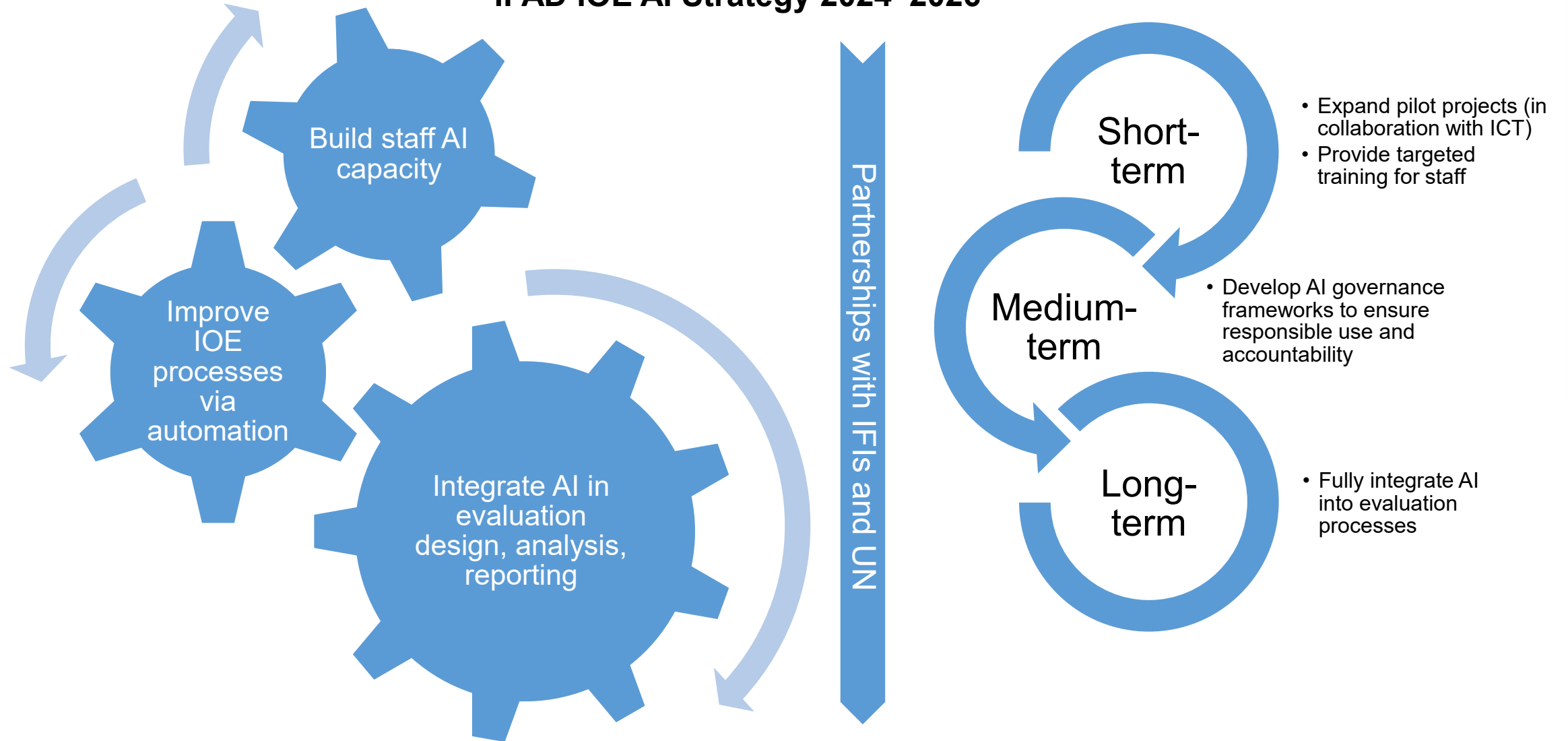
Automate repetitive tasks to free up human experts for high-level sense-making, judgment, and contextual understanding



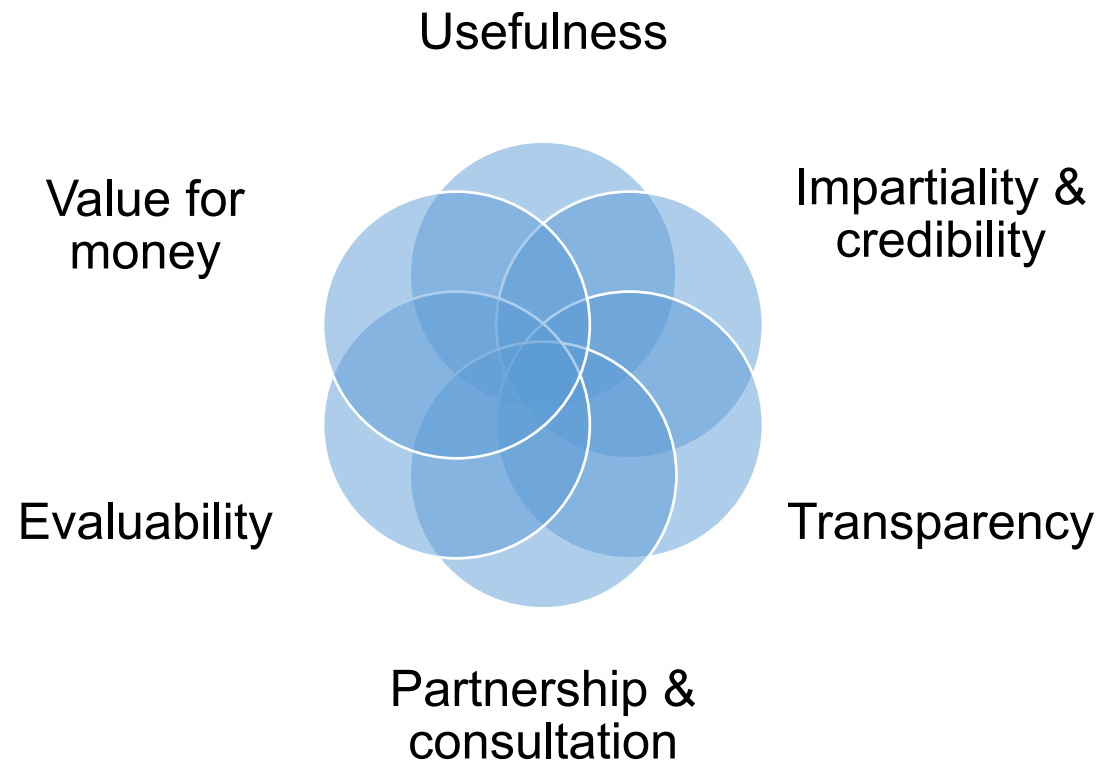
Commitment to Adapt Technology and Innovation

Aligned with IFAD's commitment on innovation and technology as per Evaluation Policy

IFAD IOE AI Strategy 2024–2026



IOE Principles Anchoring AI



[Ref- IFAD Evaluation Policy 2021](#)

The 'What': IOE Use Cases (Successes & Failures)

Use Cases: Evaluation and Beyond

Chatbot retrieval: key point extraction from HQ interview

Structured Interview analysis for country case studies: modular pipeline, synthesis across stakeholders

Document analysis: 95+ reports classified, operational parameters indexed

Triangulation: Sub-EQ matrices checked across 3+ evidence blocks

AI Powered dashboard for country operations analysis (Failed attempt for text analysis)

Chatbot for website – (ongoing)

AI powered newsletters – curated weekly evaluation/AI insights

Social media analysis – sentiment & reach tracking

Random desk allocation tool – fair workspace distribution

Smart formatting support – automated editing, consistency (failed attempt)

Report search engine – thematic indexing across corpus (Planned)

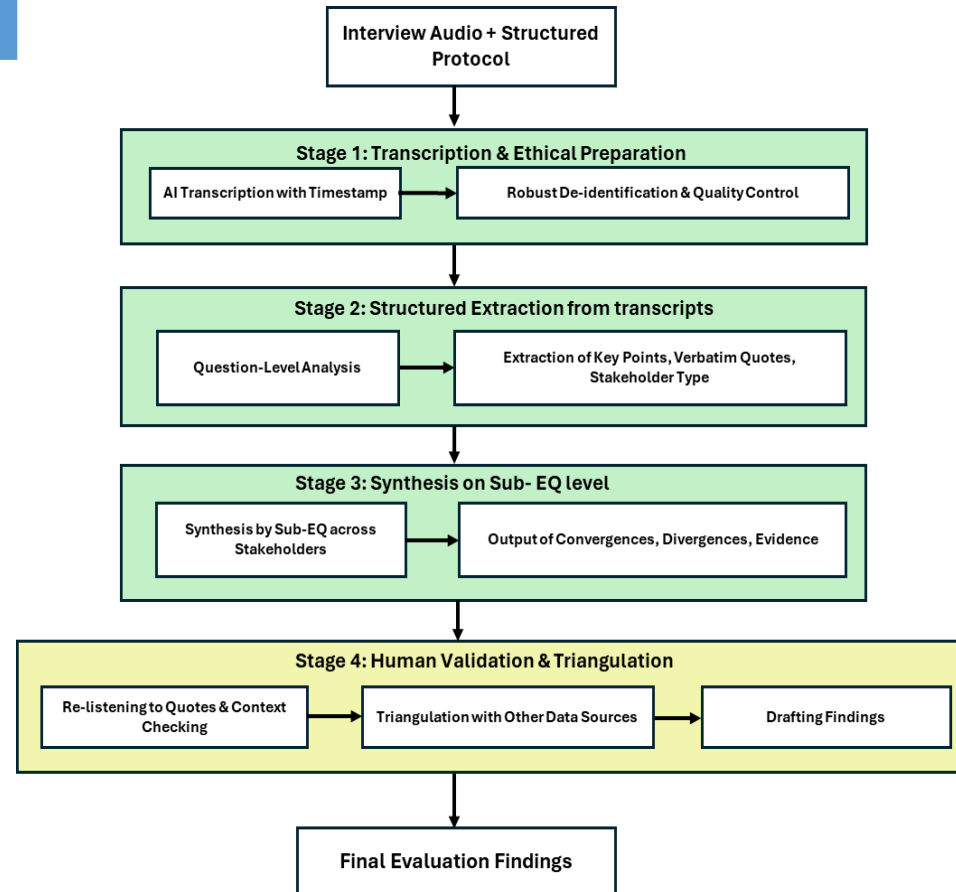
Automated meeting summaries – draft minutes from recordings

Success was scalable; failure was educational.

Overview

- CLE country case studies had well defined structured protocols, and were conducted online
- 18 -25 interviews per case study, with gov't officials, beneficiaries, partners.
- ~200,000 words of transcripts to analyze manually.
- Manual coding was time consuming, and risk of inconsistency, and loss of nuance across a vast evidence base.

Process



Legend:

- AI-assisted (with evaluator validation at each stage)
- Human-led (final validation, triangulation, and judgment)

Results

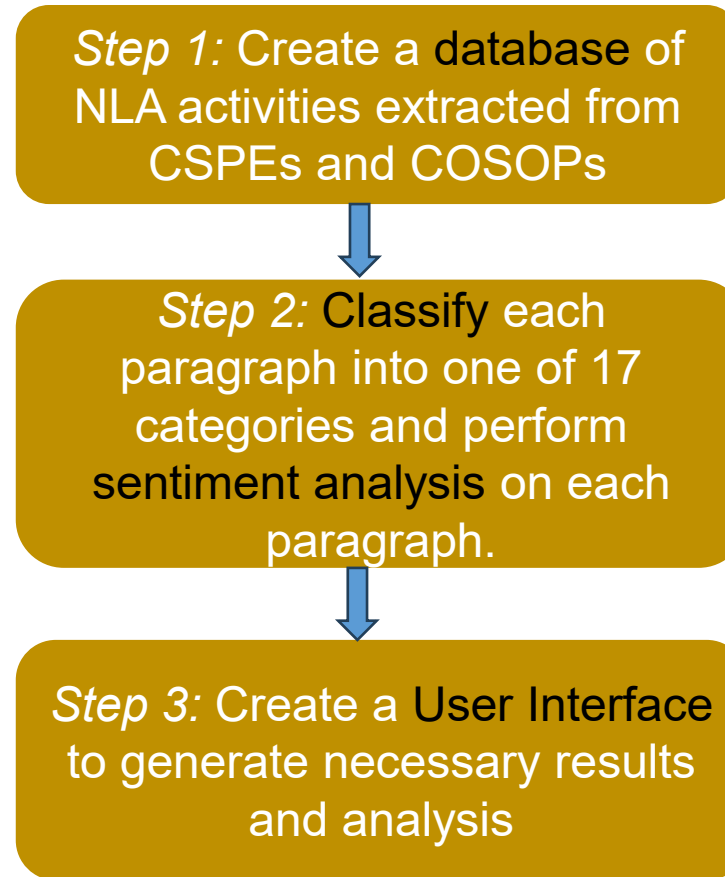
- Efficiency: Manual process: 2-3 weeks. AI-assisted: 2-3 days.
- Coverage: Every single transcript response was structured into an evidence database
- Traceability: A clear audit trail from a timestamped audio clip to a final finding.
- Evaluators focused on higher-order tasks

Overview



- Extract information on non-lending activities from 62 COSOPs and 35 CSPEs
- 17 categories of Knowledge Management, Policy Engagement, Partnership Building.

Process



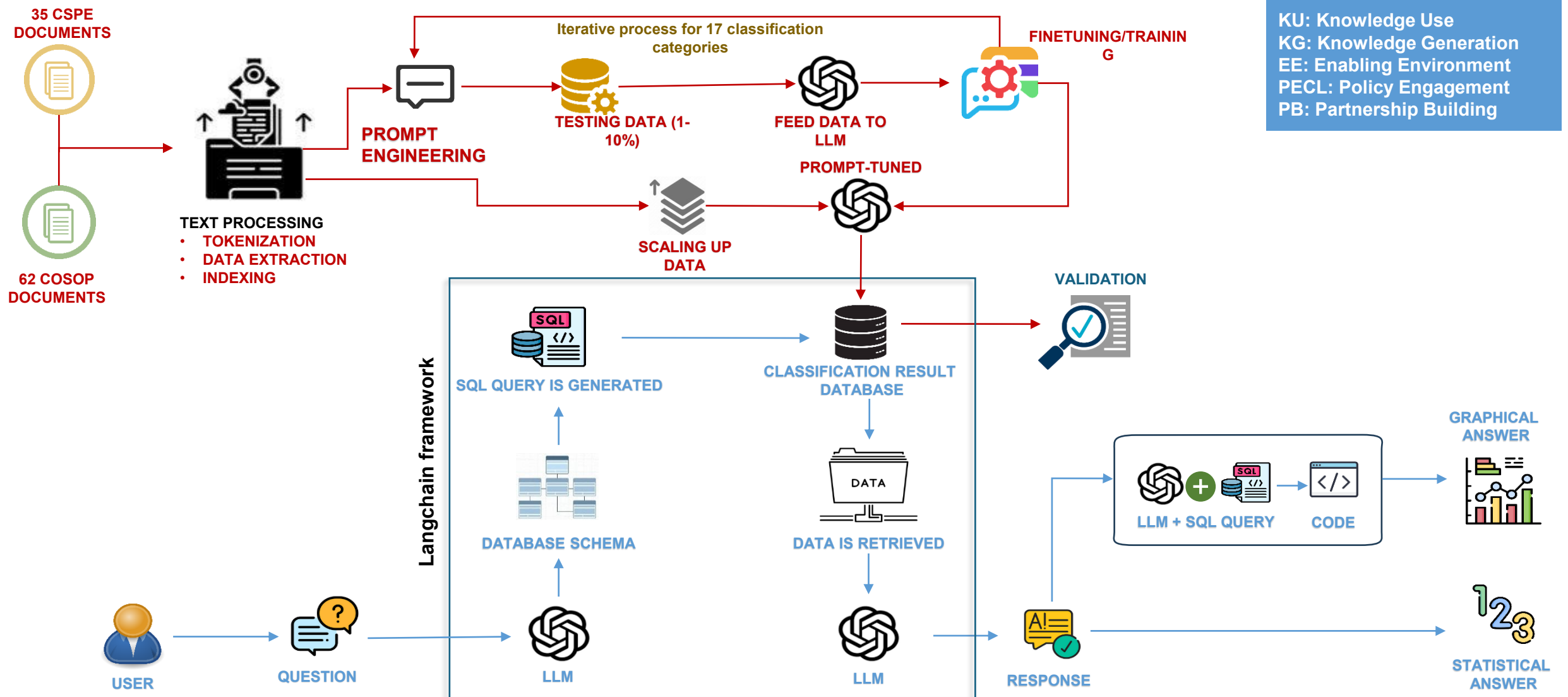
Results



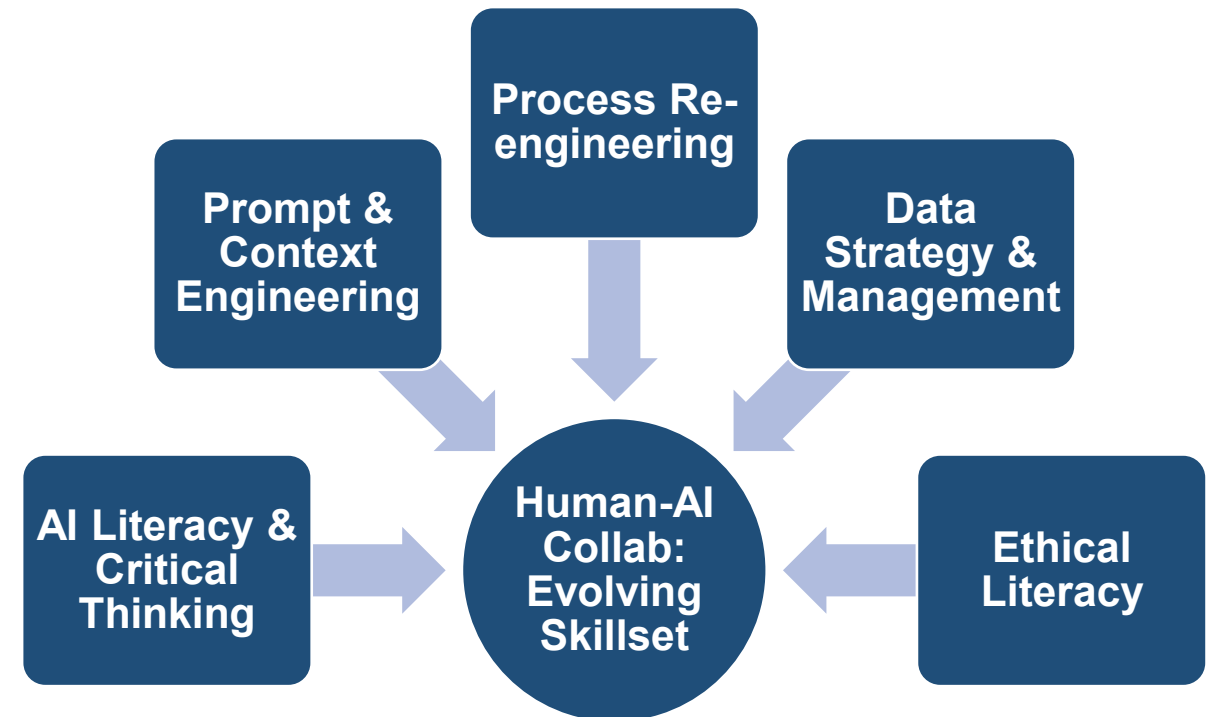
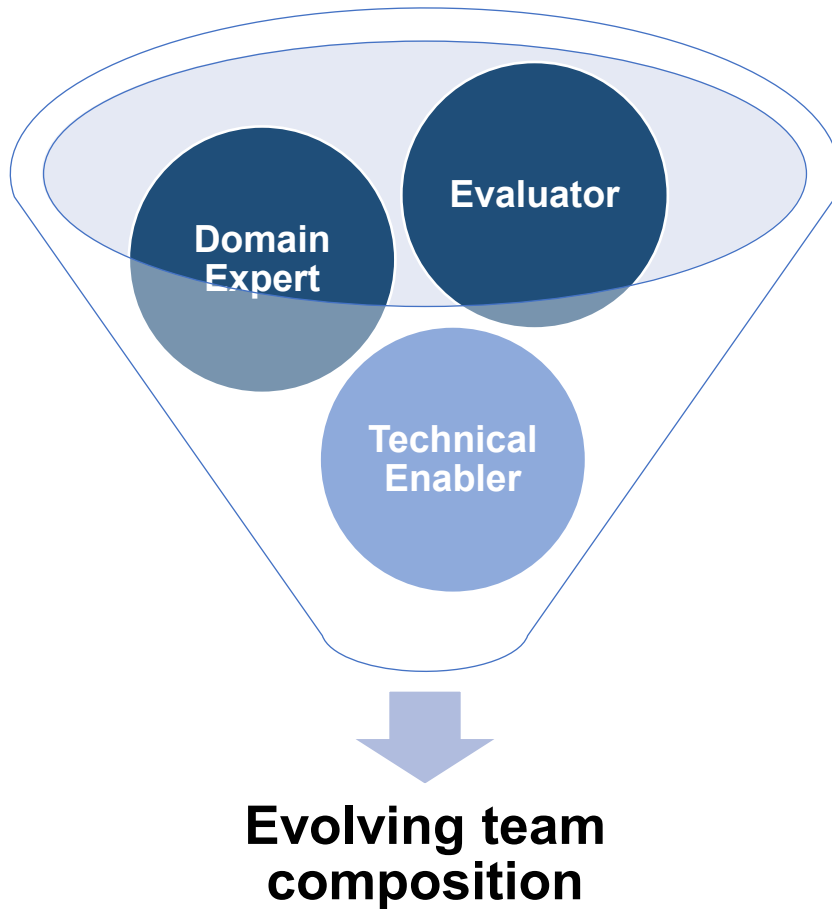
- Generate necessary graphs and tables of non-lending descriptive statistics
- 1500 Validated with F1 Score 0.82

Validation

Deepdive: Analysing non-lending activities



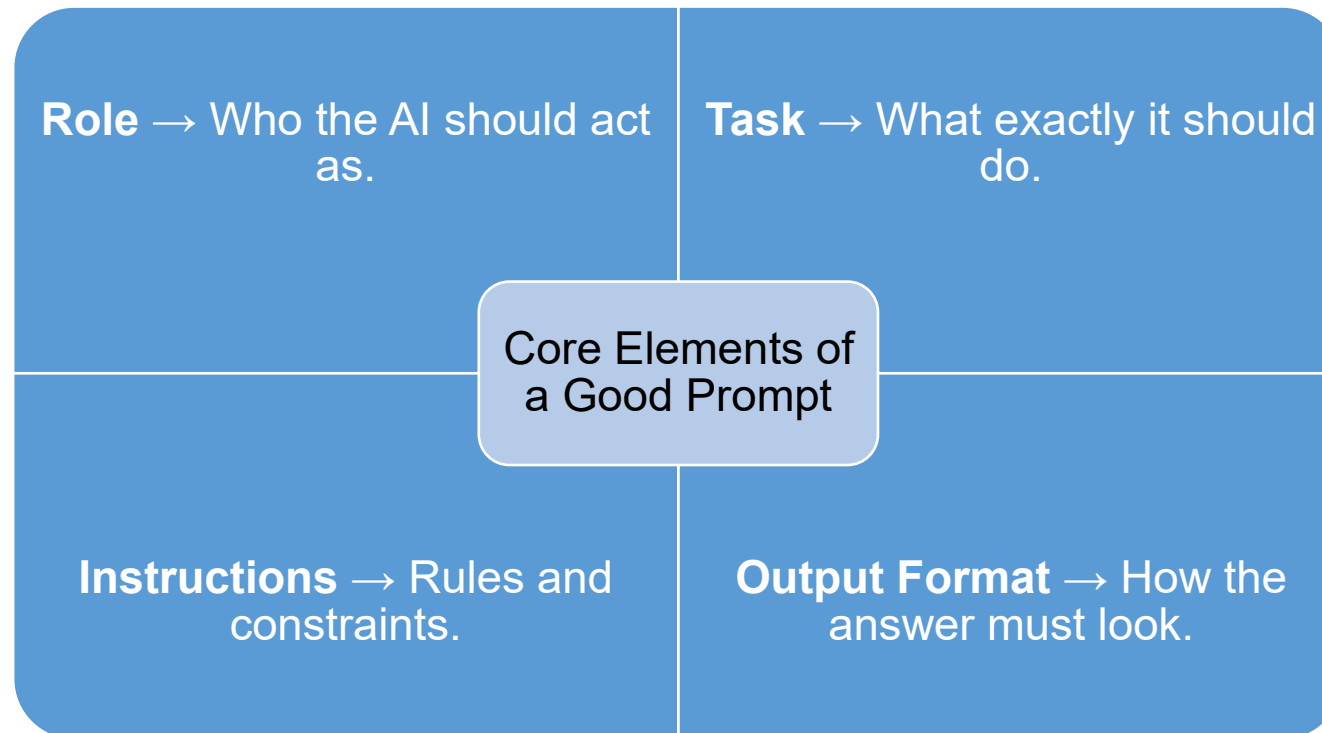
Evolving Team Composition and Evaluator's Skillset



Prompt/Context Engineering

[OpenAI \(2022–2023\). OpenAI Cookbook & Best Practices for Prompting.](#)

Prompt engineering, *Prompt engineering is the art and science of designing effective instructions for AI models so they generate accurate, reliable, and useful outputs.*





Example Paragraph: *"The project team worked closely with national stakeholders and local organizations to expand outreach to rural farmers. Regular meetings were held to coordinate inputs and share lessons, which also informed ongoing dialogue with ministries."*

✗ Bad Prompt *"Read this paragraph and tell me what kind of partnership it is."*

Output: *"This is a partnership between government, NGOs, and farmers, focused on capacity building and knowledge sharing."*



Good Prompt : **[Role]** You are a partnership specialist.

[Task] Classify the paragraph into one or more of the following categories:

1. Financial Institutions/donors
2. Rome-Based Agencies & UN Country Teams
3. Civil Society Organizations
4. Farmer Organizations
5. Research Organizations
6. Private Sector
7. Policy Engagement Partnerships
8. Knowledge Partnerships

[Instructions] Respond only with the category name(s) and number(s). If none apply, respond with [no partnership detected].

Output format: "Category (number)"

Output

- * *Farmer Organizations (4)*
- *Policy Engagement Partnerships (7)*

Types of Prompt Engineering

Zero/Few-Shot = Examples for training.

Chain-of-Thought = Improves reasoning.

Role + Context = e.g Aligns with frameworks.

Output Constraints = Makes answers usable.

Iterative Refinement = Improves reliability.

Gains



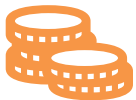
Efficiency



Consistency



Traceability



Coverage

Challenges



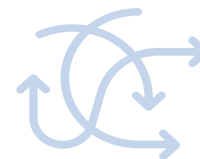
**Context-sensitive interpretation
still requires human expertise**



**Heavy validation &
benchmarking workload**



Staff adoption challenges



Nuance & ambiguity remain

Risk Category	Specific Risks	Mitigation Measures (Technical + Process)
Bias	- Data bias - Algorithmic bias	- Diverse, balanced datasets - Fairness checks & bias audits - Human validation across diverse sources
Quality & Reliability	- Mis/disinformation – - Hallucination - Validation gaps	- Retrieval-Augmented Generation (RAG) - Parameter tuning (temperature, top-p) - Fine-tuning / grounding with evaluation corpora - Human-in-the-loop validation - Cross-checking & benchmarking (RACCA, F1)
Other Risks	- Dependency on ICT infrastructure - Mistrust of technology - Data privacy & evaluation ethics	- Local/on-prem LLM hosting & backup workflows - Training & literacy for staff adoption 🛡️ - Secure/local processing, anonymization - Ethical & data-sharing guidelines

Lessons Learned



AI is most powerful in evaluator-led, question-driven workflows



Human validation is non-negotiable (cross-checking, RACCA, F1 Score triangulation)



Documentation ensures transparency, replicability, and learning



AI supports both evaluations & operations



AI outputs = inputs only, evaluators in charge



Responsible and Ethical Use (Secure, local processing & anonymization)

Thank You!



We value your time and
look forward to your ques-
tions.



Q&A

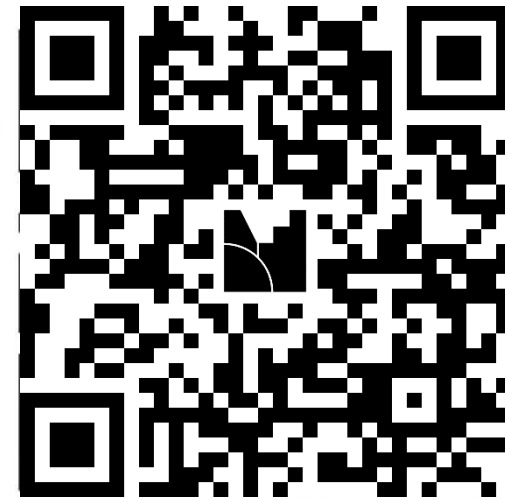
Questions?
We're here to discuss.



We'd love your feedback!

Go to www.menti.com
and enter the code

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What is key takeaway you are leaving with from today's session?

