

Brief

Scoping Study on the Use of Artificial Intelligence in Climate Change Evaluations

Introduction and Objectives

AI has the potential to improve the quality and efficiency of evaluations. It has significant value in its analytical capabilities, given the increased focus on evidence-based and data-driven decision making, and the role of data as the engine of AI. **This scoping study explores the potential of AI, assesses risks and opportunities, and provides insights into future applications in climate change evaluations.**

Methodology

- Literature review
- 19 semi-structured interviews with International Organization (IO) representatives
- Online survey of evaluation practitioners

Cautionary Notes

- AI is a broad term with diverse tools and methods
- Findings are time sensitive due to rapid AI evolution
- Ongoing supervision and adaptation required
- Study is based on peer organization experiences within the UN working in climate change

Common AI Categories

The terminology used in the AI field is complex due to the diversity of methods and approaches used. The following list contains some key terms relevant for this study:

- **Machine Learning (ML)** - focuses on algorithms and models, instructs computers to learn from large data and make predictions or decisions without being programmed for specific tasks.
- **Non-Generative AI** - analyzes and reorganizes, without being able to generate any new content.
- **Generative AI** - a system capable of creating new content out of existing information. This includes text, images, voice, or other forms of data.
- **Natural Language Processing (NLP)** - uses ML techniques to enable computers to understand, interpret, and generate language in a meaningful way. NLPs perform tasks like translation, sentiment analysis, and summarization of text. NLPs can be of both generative and non-generative in nature.
- **Large Language Models (LLMs)** - a subset of NLP, LLM models are typically trained on vast amounts datasets, which enables them to understand information and generate human-like text.

Key Findings - AI in evaluations

Current use:

There is an increasing adoption of AI use, though slower than in other fields. In climate evaluations, despite its potential, use is limited. Some common uses of AI within international organizations include:

- Document synthesis and summarization;
- Qualitative data analysis, external validity and coding;
- Evidence extraction from large datasets;
- Geospatial analysis (e.g., satellite imagery);
- Climate modeling and impact prediction; and
- Sentiment analysis of stakeholder feedback.

Opportunities:

- Enhanced efficiency and quality of evaluations
- Cost and time savings
- Better handling of complex data and systems and larger sample sizes
- Improved predictive capabilities for climate impacts

Challenges, considerations and risks:

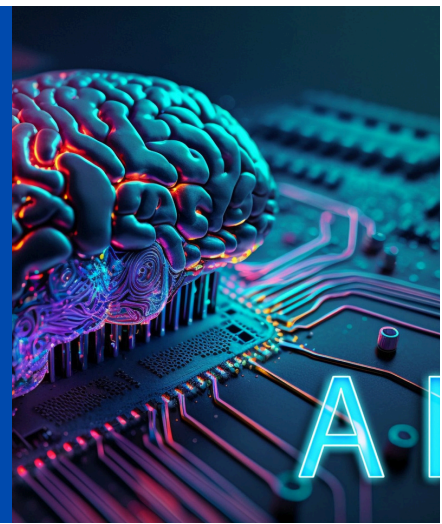
There is a strong need for human oversight and validation when considering the use of AI in evaluations. Human judgment remains crucial for context and causality. For data and technology, there are common concerns for privacy, security, bias, and quality. For generated results, there is a risk of perpetuating existing biases, interpretability issues, and “black box” challenges. There is therefore a need to balance efficiency gains of AI with result quality.

For evaluators, there is a need for capacity building in AI literacy, issues of unequal access to AI tools, and determining best practices in AI integration into evaluation processes.

Risk Mitigation:

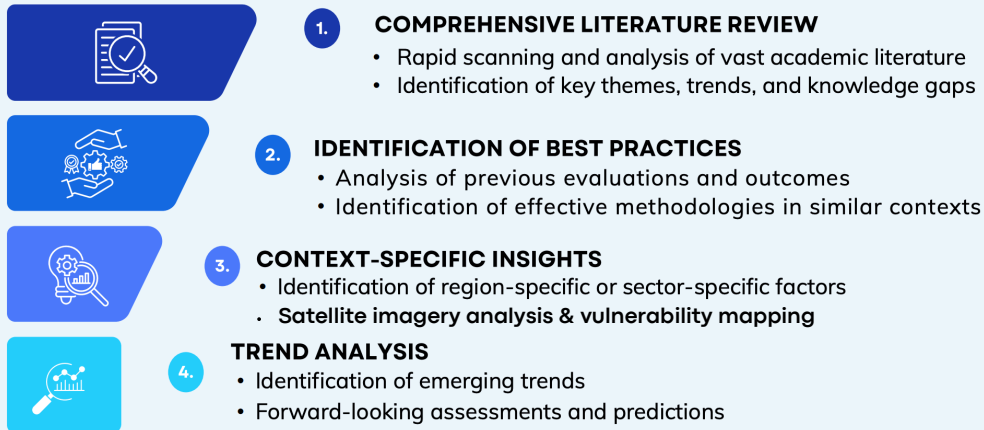
- Human-AI collaboration and oversight
- Rigorous validation processes
- Development of ethical guidelines and frameworks
- Continuous learning and adaptation
- Investment in AI literacy and capacity building

“AI can contribute to qualitative analysis across contexts. On the other hand, deeper meanings, lived experiences...currently remain challenging for AI.”

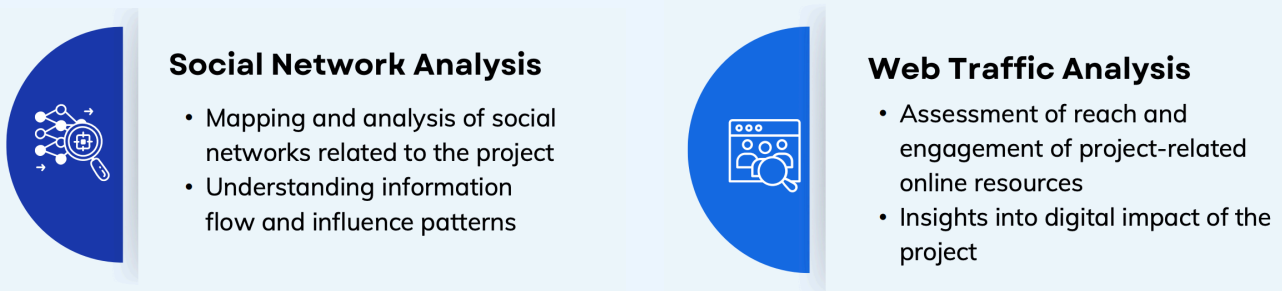


Examples of uses of AI in evaluation phases

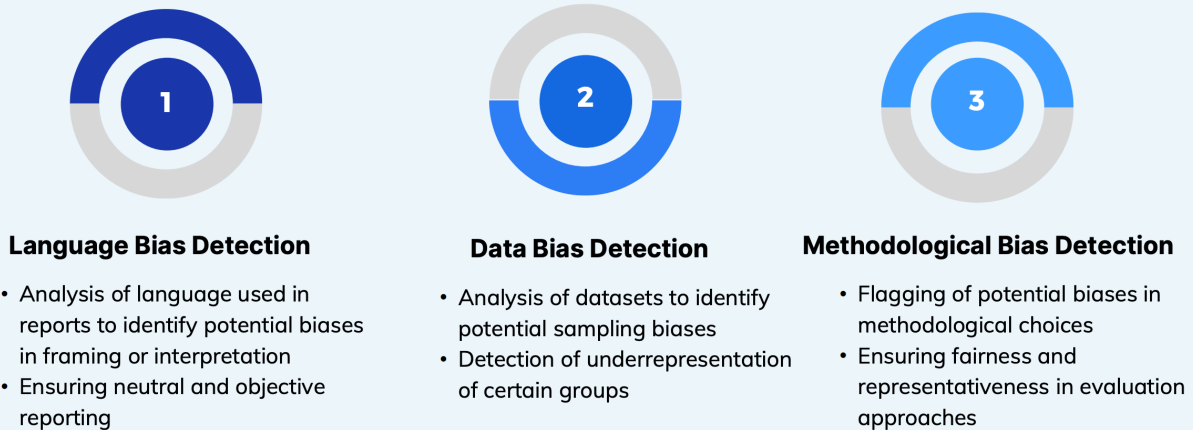
Design: *Synthesis of existing evidence & knowledge*



Implementation: *Social media and web analytics*



Quality Control: *Bias detection*



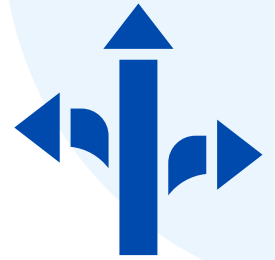
Use & Dissemination: *Synthesis of evaluation results*



Recommendations

Strategic Approach

- Adopt AI use gradually, critically, and responsibly
- Start with simple applications, then progress to complex tasks
- Develop tailored AI applications for evaluation needs
- Foster a culture of experimentation and innovation
- Continuously assess the impact of AI and its value-added



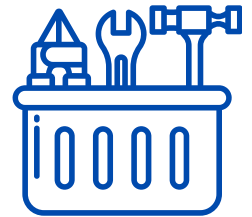
Governance and ethics

- Develop AI governance frameworks and ethical guidelines
- Establish protocols for data privacy and security
- Create risk assessment procedures for AI projects
- Ensure transparency in AI use within evaluations
- Conduct regular ethical reviews of AI systems and outputs



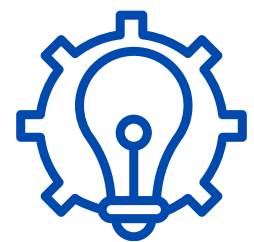
Capacity building

- Invest in AI skills and training for evaluation staff
- Promote collaboration between evaluators and AI experts
- Organize workshops and knowledge-sharing sessions
- Develop AI literacy across all levels of organization
- Create a core team of “AI champions”



Implementation

- Conduct AI readiness assessments
- Implement pilot projects with clear objectives
- Develop AI-assisted evaluation toolkits
- Establish validation frameworks for AI outputs
- Create feedback loops for continuous improvement



About the study: By Elinor Bajraktari, independent consultant, and commissioned jointly by the Technical Evaluation Reference Group of the Adaptation Fund (AF-TERG), Climate Investment Funds (CIFs) Evaluation and Learning Initiative, Global Environmental Facility Independent Evaluation Office (GEF-IEO), and Green Climate Fund Independent Evaluation Unit (GCF IEU).