

EXPLAIN: ESSENTIAL BRIEFINGS FOR HUMANITARIAN DECISION-MAKERS AI IN THE HUMANITARIAN SECTOR

When discussing Al in the humanitarian sector, it is crucial to distinguish between its potential value and its proven impact. While generative Al (GenAl) is widely accessible and hyped for its efficiency gains, a persistent knowledge gap makes it difficult to assess its real benefits. Humanitarian decision-makers need critical literacy to navigate Al's actual impact, deployment challenges, and budgeting requirements.

Humanitarians must recognise both the opportunities and risks of Al. Traditional Al analyses and automates, while generative Al (GenAl) creates new content based on trained data. GenAl tools like ChatGPT and Gemini rely on large language models (LLMs) trained on vast public datasets, aiding humanitarian tasks like information synthesis and report generation. However, concerns around bias, transparency, reliability, and ethical risks—such as misinformation and nonconsensual data use—pose challenges for humanitarian applications.

Generative Al in the humanitarian sector is emerging, but not yet systematic. While discussions on Al often focus on generative Al (GenAl), non-generative Al tools like machine learning and predictive analytics remain more prevalent in automating key processes. Most humanitarian use of GenAl is limited to <u>backend tasks</u>, with large-scale adoption hindered by <u>uncertainty</u>, risk concerns, and a lack of technical expertise.

Humanitarians must define the use

of Al tools in programming to assess their risks, benefits, and relevance.

- GenAl improves efficiency in everyday tasks like drafting emails, summarising meetings, and editing reports, though its adoption in humanitarian work remains informal and difficult to track.
- 2. Al is increasingly integrated into institutional workflows, enhancing internal processes and automating humanitarian tasks to improve efficiency at scale.
 - Al enhances internal knowledge management by cleaning data, making <u>databases queryable</u>, centralising information, and streamlining task management, while also improving <u>supply chain</u> <u>management</u> through real-time data integration and tracking.
 - Big data analytics support MEAL tasks by extracting insights, synthesising data, processing surveys and feedback, generating metadata, and coding lessons learned, as demonstrated by <u>Catholic Relief</u> <u>Services' use of machine learning</u> in Malawi to assess food insecurity risk.
 - Al aids decision-making in humanitarian settings by improving population density estimates for aid distribution, using semantic segmentation to assess <u>building</u> damage, supporting vaccination

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programs, and assisting in <u>medical</u> <u>applications</u>, with the <u>European</u> <u>Crisis Management Laboratory</u> testing Large Language Models (LLMs) for extracting news and compiling time-sensitive risk data.

- 3. Al tools enhance community engagement by personalising interactions through platforms like WhatsApp and Facebook, with <u>chatbots</u> varying from simple decision trees to more advanced NLP- and GenAlpowered models. However, challenges in accuracy, cultural appropriateness, and oversight—such as those seen in <u>IRC's Signpost initiative</u>—highlight trade-offs in Al sophistication, while <u>sentiment analysis</u> helps process community feedback more efficiently.
- 4. Al is expanding humanitarian capabilities by enabling tasks previously limited by cost or resources, such as <u>Al-driven family reunification</u> and anticipatory approaches predicting displacement. While models like the <u>Danish Refugee Council's AHEAD</u> and <u>WFP's VAM</u> use machine learning to forecast displacement and food shortages, these applications involve high-risk factors, including predictive analytics, sensitive data, and potential unintended consequences.

Al tools pose risks such as misinformation, bias, opacity, and privacy concerns, while also threatening the humanitarian mission by undermining the 'do no harm' principle, exacerbated by the lack of common standards.

• Data gaps and misinformation: Poorquality data leads to inaccuracies, mistranslations, and misinterpreted nuances, with <u>AI tools often trained</u> in data-rich languages like English, leaving humanitarian contexts underrepresented and <u>increasing</u> risks to operations and communities.

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- Bias and discrimination: Al outputs reflect biases from dominant datasets, and the lack of community input and local knowledge heightens the risk of inaccurate, irrelevant, or discriminatory decisions.
- Reliability challenges: GenAl tools frequently generate false information ("hallucinations"), while model drift degrades Al accuracy over time, leading to misinformation, service denials, or flawed reports.
- Privatisation and vendor lockin: Dependence on private Al providers risks undermining humanitarian neutrality and increasing reliance on proprietary systems with limited control.
- **Privacy and consent risks:** Al tools handling sensitive data expose communities to security breaches, while adoption pressures may lead to exploitative pilot programs where informed consent is fragile.
- Opacity and accountability: Al decision-making is often opaque, limiting oversight and making it difficult for individuals to challenge service denials or verification failures, especially with proprietary models.
- Human rights implications: Al's energy demands strain infrastructure, its data-labelling process is linked to labour exploitation, and Al-driven misinformation complicates humanitarian efforts.

The rise of GenAl has raised the stakes for humanitarians, making shared standards more urgent than ever. The rapid rise of GenAl has intensified the need for shared standards, as its adoption in the humanitarian sector grows without clear safeguards. Previously used for specialised tasks, Al is now widely accessible, leading to increased experimentation, but without a common framework, its use remains informal and fragmented.

Rigorous research is needed to assess Al's utility and costs in humanitarian operations, with clear boundaries for GenAl use, as most tools are not designed for humanitarian needs.

- Stronger research on GenAl in humanitarian contexts: Humanitarians need applied research and critical literacy to assess GenAl's risks, including data privacy concerns and the opacity of Al-driven decisions.
- Understanding costs, skills, and risks: Organisations must evaluate whether the financial and human resources required for GenAl development and testing are justified, especially given funding constraints and uncertain returns.
- Supporting staff in responsible Al use: Clear guidance and training are essential to help humanitarian staff critically assess Al-generated content, protect sensitive data, and understand Al's limitations.

Humanitarians need holistic assessment approaches to fully capture Al's ethical risks, ensuring informed and responsible adoption.

• Keep communities at the centre: Al adoption must prioritise <u>participatory design</u>, community consultation, and inclusion to ensure accessibility and relevance.

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- Establish clear standards and Policies: Humanitarian actors should define acceptable AI use by referencing established standards and creating policies that set ethical principles, testing protocols, and boundaries.
- Al is not always the answer: <u>Decision-makers must assess whether Al</u> <u>genuinely adds value</u> or if simpler, safer solutions are more appropriate to avoid unnecessary complexity.
- Understand and assess risk: Organisations must continuously evaluate AI tools, scrutinise claims, and ensure they have the capacity to audit evolving risk profiles.
- Account for resource demands: Al adoption requires investment in data collection, expertise, risk assessments, and maintenance, which, if overlooked, can lead to unsustainable implementation.
- Share knowledge and learnings: Openly sharing AI experiences, challenges, and validated use cases can help close the sector's knowledge gap and promote responsible adoption.
- Build, buy, or borrow?: Organisations must weigh security, customisation, and ethical trade-offs when deciding whether to develop, purchase, or adapt Al solutions.
- Develop a humanitarian-specific Al ecosystem: Tailoring Al to humanitarian needs, such as using Small Language Models trained on verified data, can improve privacy, governance, and resource efficiency.

Interest in GenAl is growing amid tight budgets, but its adoption must be driven by community needs rather than cost-cutting pressures. Humanitarians must critically assess claims about Al's efficiency while considering risks, ethical concerns, and the impact on crisis-affected communities. Safe, responsible Al use depends not just on the technology itself but on the systems that ensure its ethical implementation, reinforcing the importance of past lessons on innovation and the "do no harm" principle.

This quick read delivers the essentials from ALNAP's full briefing on AI in the humanitarian sector.

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