

### AI in Africa: The state and needs of the ecosystem

Diagnostic and solution set for data

March 2024



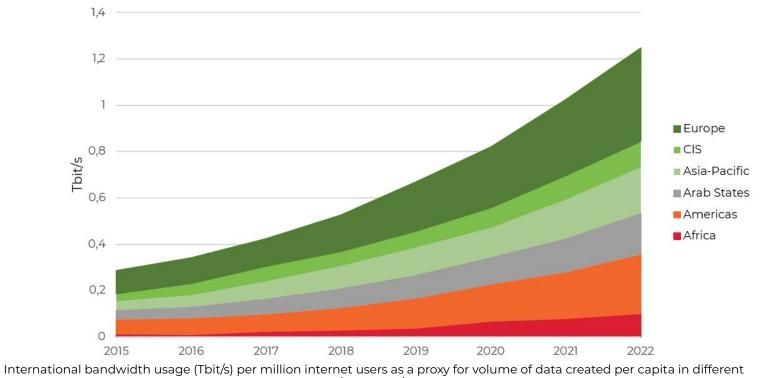


#### WHAT IS COVERED IN THIS PACK

- Data challenges across the continent
- 2. Channels for plugging the data gaps on the continent
- 3. Stylised case studies with associated lessons learnt and high level statistics on:
  - a. Developing tangible use cases to scale sector-related datasets
  - b. Using public data to achieve a quick win
  - c. Mapping non-standard smallholder farms to improve decision making
- 4. Quantifying the gap in data and associated investment ask
- 5. Suggested areas for intervention

#### In a world of increasing data generation, Africa is a laggard

Africa lags behind in the number of citizens connected to the internet, and by proxy able to generate data.



regions over time.

#### One of the largest gaps in the availability of data is in language

'Brute force' alone is unlikely to achieve parity in data generation in African languages compared to Anglophone, given the extent of the data imbalance.

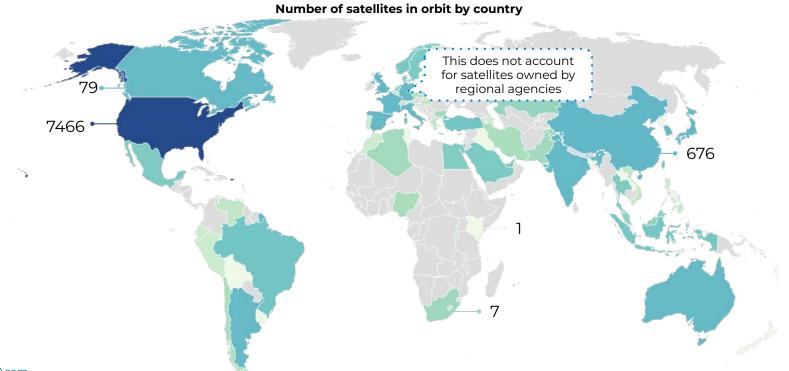
Global	English	52.60%
Benchmarks Top African Languages	Hindi	0.1%
	Afrikaans	0.003%
	Twi	0.00195%
	Swahili	0.00135%
	Malagasy	0.00022%
	Bambara	0.00025%
	Venda	0.000115%
	Hausa	0.00011%
Average with other African languages*		0.000999%
Sum of African Languages		0.01999%

Language (Nov '23) % Share of Internet content in local language

\* Afrikaans, Twi, Swahili, Bambara, Malagasy, Hausa, Venda, Haitian, Haitian Creole, Igbo, Luba-Katanga, Ndonga, Rundi, Tokelau, Tswana, Akan, Chichewa, Chewa, Nyanja, Fulah, Ganda, Masai

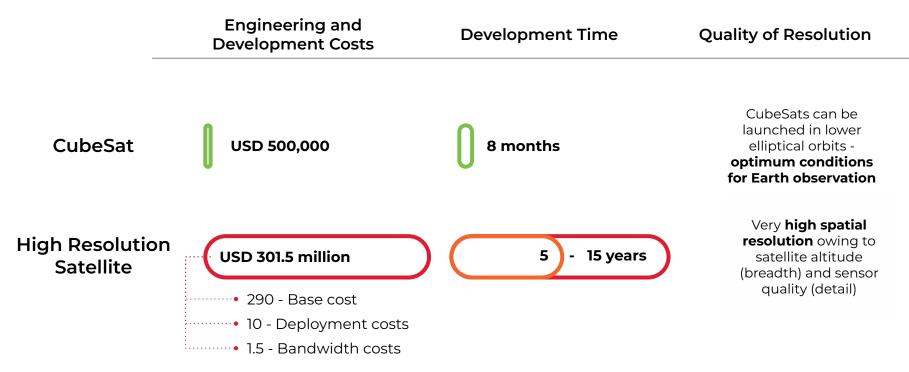
#### Majority of African countries are without a satellite in orbit

High spatial resolution remote sensing enables real-time decision-making. However, the cost of the sophisticated sensors required for such quality imagery limits their accessibility, particularly in Africa.



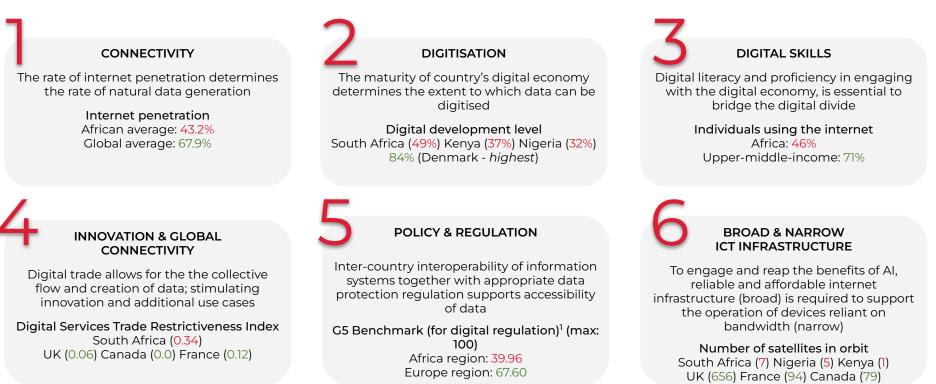
## Before CubeSats, earth observation data required a compromise in data - either occasional high spatial resolution, or *frequent* low resolution

A CubeSat is a miniaturized satellite that is used to collect data and enable surveillance and monitory and communications. It is a **low-cost, lightweight option with a standard design** that provides for high spatial resolution data collection.



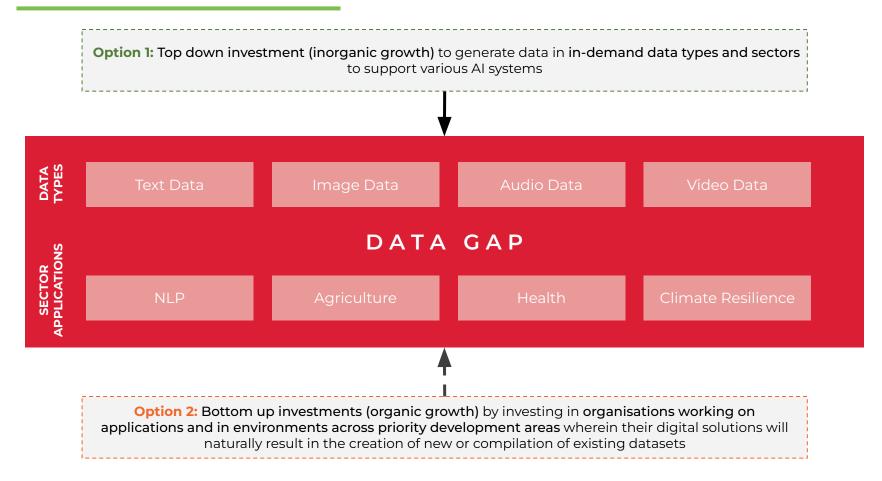
#### Data gaps result from constraints in the underlying environment

These constraints impact the development or scaling of business or government sector-related use cases; the quality or completeness of datasets; and the availability of data to build solutions in vernacular.



<sup>1</sup>Measures the state of collaborative digital regulation

#### Two main channels exist for plugging the data gaps on the continent



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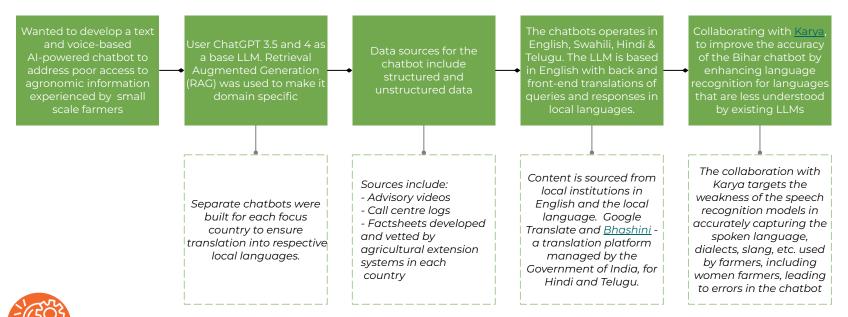
	<b>Option 1: Top down investment to generate inorganic growth</b> of datasets through the <b>scaling of</b> in-demand data types to support various AI systems						
DATA TYPES	Text Data	Image Data	Audio Data	Video Data			

Stimulating organic growth of datasets is likely to be more attractive and sustainable; but, will take longer given the nascency of the ecosystem and scale of the underlying constraints

SECTOR APPLICATIONS	NLP	Agriculture	Health	Climate Resilience	
	Option 2: Stimulating organic growth of datasets by investing organisations working on sector applications across priority economic development sectors wherein their digital solutions will naturally result in the creation of new or compilation of existing datasets				

#### Digital Green: Developing tangible use cases to scale sectorrelated datasets

Digital Green's Farmer.CHAT solution leveraged multiple data types, technology services providers and open source platforms to build an AI-powered chatbot that supports agricultural extension workers.

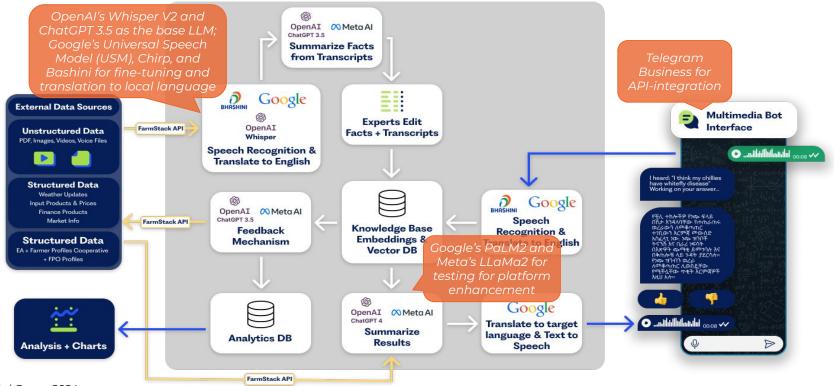


The **successful business use case** of Farmer.CHAT has resulted in significant demand from public sector partners. Digital Green has raised **USD 30 million** to support the development and rollout of similar AI-powered agronomic chatbots for agricultural ministries in India, Kenya and Ethiopia - with the aim of reaching over 220 million farmers in each country.

Source: Digital Green. 2024

## The success of Farmer.CHAT is contingent on important partnerships with Technology Service Providers

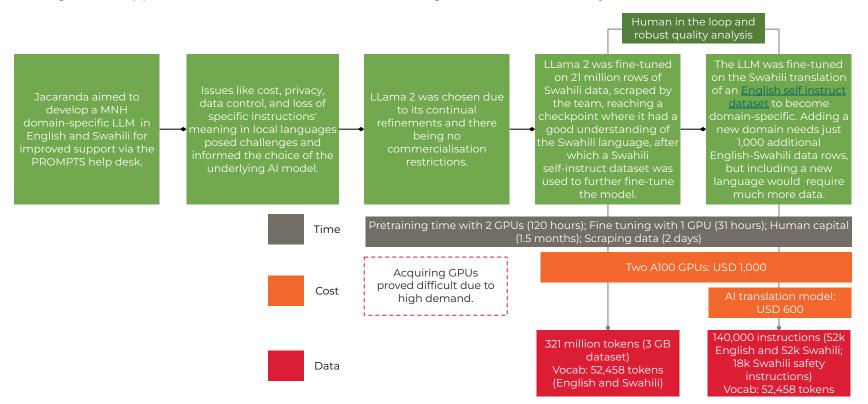
Digital Green's Farmer.CHAT solution leveraged multiple data types, technology services providers and open source platforms to build an AI-powered chatbot that supports agricultural extension workers. The diagram below demonstrates the complexity of the solution, and the numerous areas where Big Tech can contribute



Source: Digital Green. 2024

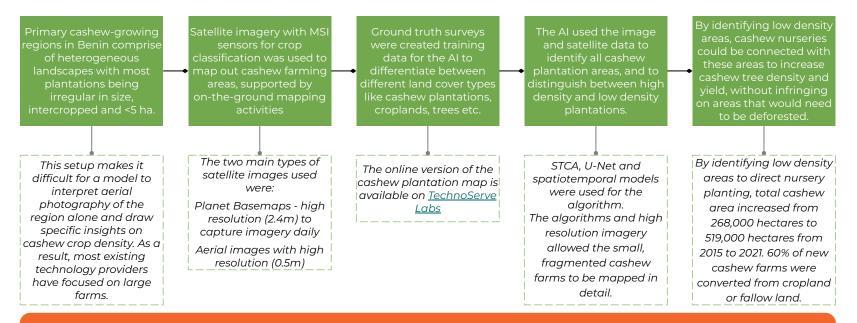
#### Jacaranda Health: using public data to achieve a quick win

Jacaranda Health developed UlizaLlama to enhance accuracy and context-specificity of its automated responses on the maternal and newborn health (MNH) digital platform, PROMPTS. The project showcases the feasibility of scaling similar applications and contributes to broadening the African NLP ecosystem.



# TechnoServe Labs: mapping non-standard smallholder farms to improve decision making

Accurate crop mapping is vital for securing livelihoods, yet current models struggle to identify nonstandard smallholder farms predominant in Africa. According to Technoserve, **USD 1.5 million, coupled with 5-7 software developers**, would be the **investment required per crop type**.



Access to GIS and satellite imagery will become increasingly important in the agriculture and climate resilience sectors, particularly as <u>regulations</u> aim to protect deforestation efforts present new challenges for smallholder farmers to map out and certify their farming areas to ensure compliance with export standards.

#### Quantifying the gap in data and associated investment ask

The investment gap for data is estimated using funding amounts provided to grantees in priority sectors by Lacuna Fund. Funding provided by Lacuna aims to address the gap in the lack of openly accessible datasets in African languages.

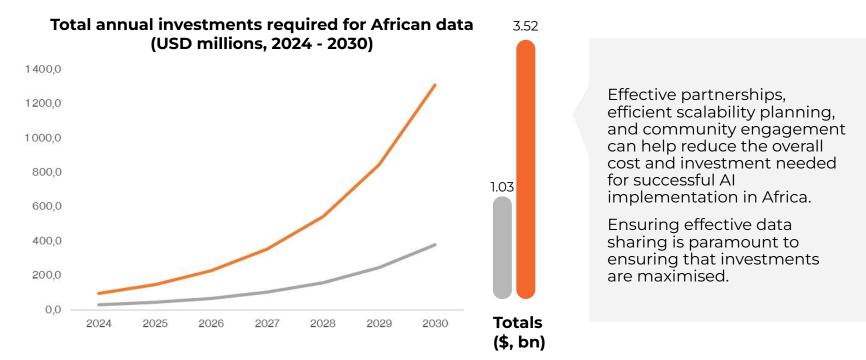


Source: Lacuna Fund\* (\*Compiled in 2023 for Grantee Cohorts in 2020-2022)

See Methodology <u>here</u>

#### Quantifying the gap in data and associated investment ask

Quantifying the data gap across the continent for all sectors is a complex task - it can be sized in orders of magnitude of error. The estimated investment required ranges from just over **\$1 billion to \$3.5 billion**. Based on Lacuna Fund data. Agricultural and Climate data make the bulk (64%) of necessary investments.

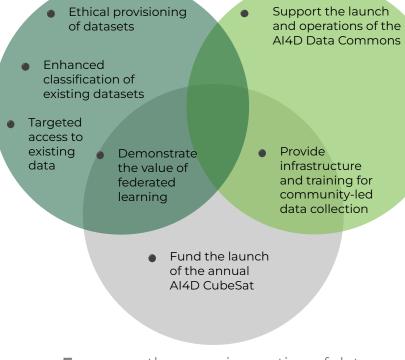


Source: Lacuna Fund: Compiled in 2023 for Grantee Cohorts in 2020-2022

### WHERE TO INTERVENE

#### Data Quick Wins

**Scale up** methods of unlocking safe access to new sources of data



**Scale out** necessary infrastructure to support AI4D Data Commons and AI Language Hub

Empower the organic creation of data

# Support the launch and operations of the Al4D Data Commons

Data sharing and integration can have a profound impact on private and public sector applications. However, AI use cases that demonstrate the effective integration of data are limited due to data scarcity and limited access to essential resources like GPUs or licensed datasets.

The launch of the AI4D Data Commons can provide a central point for data collection and sharing. The AI4D Data Commons would be a cloud-based platform that enables community members to submit, analyse, store, and share data; assisting with matching research requests with available datasets.

Big Tech can invest in a number of ways. Firstly, Google Cloud could provide free storage services. Secondly, through the provision of translation engines such as Google's Chirp to expand the usability of LLMs into different African languages. Thirdly, by allocating more training hours to fine tune existing models like USM (Google AI) and MMS (META) using African language data.



# Enhancing existing data collection processes

Limited access to data at a community-level can hinder accurate representation in datasets and stifle the development of community-level representative datasets.

Community-Led Monitoring (CLM) is a participatory approach to collecting, analysing and using data, and involves community members collecting data from their community across healthcare, education, clean water and other essential infrastructure priorities.

Internet Service Providers like MTN or Safaricom can provide CLM leaders with ICT infrastructure (mobile devices, tablets, data bundles) and digital literacy onboarding to allow these community members to capture data digitally. Players like Opera or Google can provide access to online data collection tools such as survey management platforms to execute data collection processes. This data can be stored on the AI4D Data Commons, prompting further research.

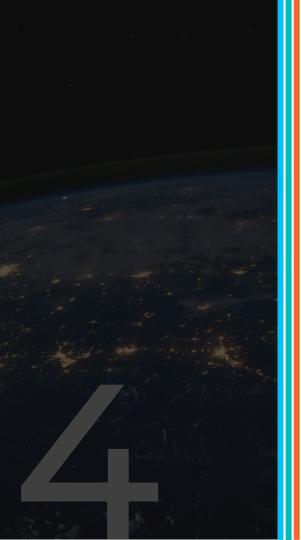


# Unlocking safe access to existing and new sources of data

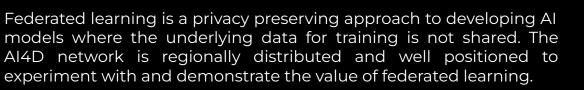
The scarcity of high-quality and diverse text data in African languages hinders the development of effective African AI and NLP models.

Enhanced data classification, as offered by Google's Dataset Search and potentially Amazon's AWS Data Exchange, can categorise and index datasets as 'relevant to Africa' making them discoverable through keyword searches. Players like Google and Uber can also offer existing support services like help desks, and FAQs in key African languages, generating valuable, machine readable data in African languages.

ISPs like Vodafone, Telefonica, Airtel, and others can enhance local language NLP data by ethically making anonymised call center data open source to the AI4D Data Commons. Organisations like GSMA can also advocate for its members to provide similar anonymised data available.

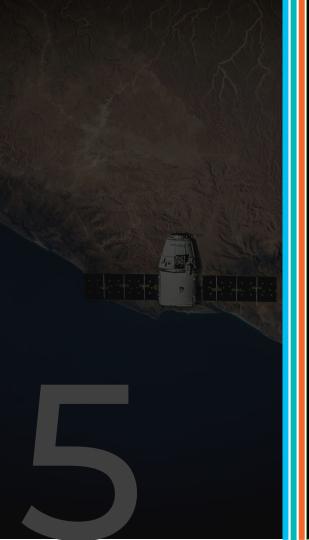


# Demonstrating the value of federated learning



Big Tech firms such as MNOs can contribute to a federated learning Proof of Concept (PoC). They can provide secure data access and infrastructure for select AI4D innovation lab members to experiment with and develop federated learning models in areas such as credit extension, customer support, or others.

These PoC's will demonstrate the opportunity of federated learning to participant firms, positively impact Africans should the PoC be well designed ,and provide the Al4D network with an opportunity to 'specialise' in an area that has impact and that they are well positioned to lead on.



# Addressing barriers preventing access to quality data

Limited access to high-resolution GIS or satellite imagery due to funding constraints can lead to incomplete datasets, especially in sectors like agriculture. Smallholder farms' undocumented perimeters inhibit the feasibility of building data-driven solutions to their challenges, leading to reliance on drone technology for data collection, which is expensive and lacks continuous data access.

The launch of four annual AI4D CubeSats, costing around \$500,000 and offering faster deployment and high-resolution imagery, could enable continuous and up-to-date monitoring of expansive regions, supporting accurate national planning decisions for sectors including agriculture, water and sanitation. Further, overlaying this data with additional information including weather forecasts can allow for more proactive climate change policy. The data collected from the CubeSat can be accessed via the AI4D Data Commons, and stored similarly.

## APPENDIX

#### Slide 16: Quantifying the gap in data and associated investment ask (1/2)

- Total amount funded, shortlisted and requested is provided by Lacuna Fund for 2020-2022 grantees (see table below, average annual investment by sector)
- Lower band is calculated as *Total Amount Funded* subtracted from *Total Amount Shortlisted* (\$7.6 million in 2024)
- Upper band is calculated as *Total Amount Funded* subtracted from *Total Amount Requested* (\$26.17 in 2024)
- These bands are then multiplied by the average annual growth rate of AI startups (55%) and the calculated 3.6 multiplier.
- This is summed with the annual investments in a CubeSat per region.

	Agri	NLP	Equity and Health	Climate	Total (millions)
Funded	0,73	0,70	0,57	1,17	3,17
Shortlisted	4,50	2,20	1,17	2,87	10,73
Requested	12,00	7,67	3,00	6,67	29,33

#### Slide 16: Quantifying the gap in data and associated investment ask (2/2)

• This multiplier figure is calculated as the number of additional key donor organisations<sup>1</sup> providing grant funding for datasets multiplied by the weighted average of the number of priority sectors<sup>2</sup>



<sup>1.</sup> This includes: <u>BMFG Grant Challenge for "Catalyzing Equitable AI Use"</u>; The <u>Rockefeller Foundation</u>; <u>Science Granting Councils Initiative in</u> <u>sub-Saharan Africa</u>; <u>UNICEF</u>; <u>GIZ</u>; and the <u>Wellcome Trust</u>

<sup>2.</sup> These sectors include: NLP; Agriculture; Health and Equity; Climate Resilience; Education and Financial Inclusion