

# Literature Review on Data Gaps in Africa

(Dataverse Africa Internship Program – Data Gap Initiative)

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### 1. Introduction

Data plays a crucial role in driving effective policies, interventions, and decision-making processes, particularly in developing regions like Africa. A **data gap** refers to "the absence or inadequacy of data needed to inform decision-making processes" (Dell, 2023). These gaps often result from issues such as incomplete, outdated, or inaccessible data, as well as the lack of structured data collection mechanisms (Hassan, 2021). The absence of reliable data prevents accurate assessments of critical needs and hinders the implementation of policies designed to address pressing challenges.

Significant gaps in data availability, accessibility, and quality persist across various sectors in Africa, limiting progress in areas such as public health, food security, education quality, and infrastructure development (Adebayo et al., 2019). Addressing these data gaps is crucial for enabling evidence-based decision-making, fostering sustainable economic growth, and ensuring effective policy implementation (UNICEF, 2019; WHO, 2020).

This literature review examines data gaps in **healthcare**, **education**, **agriculture**, **and infrastructure**, highlighting both sector-specific challenges and potential strategies for bridging these gaps. The review also evaluates successful case studies from other regions and provides recommendations for Africa's data improvement initiatives.

# 2. Literature Review: Data Gaps in Key Sectors Of Africa

#### 2.1 Healthcare Sector

The healthcare sector in Africa faces significant data limitations that affect planning, decision-making, and service delivery (Kofi & Mensah, 2021). While developed nations benefit from Electronic Health Records (EHRs) and AI-driven analytics, African countries struggle with fragmented data systems, inadequate infrastructure, and workforce shortages (WHO Africa, 2019).

#### **Key Healthcare Data Gaps**

- 1. **Electronic Health Records (EHRs):** Many African countries lack standardized digital health record systems, relying instead on paper-based documentation, which leads to poor interoperability and inefficiencies (Muniu, 2024). The lack of data integration among hospitals, clinics, and health ministries further impairs decision-making processes.
- 2. **Patient Outcome Data:** Incomplete or unreliable tracking mechanisms prevent healthcare providers from assessing treatment effectiveness and long-term health outcomes (Westerink et al., 2023). The absence of uniform reporting standards results in inconsistent healthcare evaluations across regions.

- 3. **Disease Surveillance:** Inadequate national health databases hinder effective disease tracking, making it difficult to respond to outbreaks such as Ebola and COVID-19 (Oleribe et al., 2019; Chetty, 2024). Many African countries lack real-time epidemiological monitoring systems, which delays response efforts.
- 4. **Healthcare Workforce and Infrastructure:** Limited data on personnel distribution and hospital capacities contributes to staffing shortages and resource misallocation (Ismaila, 2023). A lack of integrated workforce management systems leads to inefficient deployment of health professionals.

**African Case Study:** Rwanda's AI-powered health monitoring platforms have improved disease tracking and patient management (Twahirwa Rutikanga, 2022).

#### 2.2 Education Sector

The lack of comprehensive educational data limits efforts to improve student performance, teacher quality, and infrastructure planning (Adjadeh, 2024). Without accurate data, policymakers struggle to allocate resources efficiently and improve education outcomes.

#### **Key Education Data Gaps**

- 1. **Student Performance Tracking:** Many African countries do not have standardized assessments, making it difficult to measure learning outcomes (World Economic Forum, 2024). The absence of longitudinal data on student progress hinders efforts to develop targeted educational policies.
- 2. **Enrollment and Dropout Rates:** Inconsistent or incomplete birth registration data makes tracking student progression challenging (Mo Ibrahim Foundation, 2024). Rural areas are particularly affected due to logistical difficulties in data collection.
- 3. **School Infrastructure Data:** Many education systems lack detailed records on classroom availability, internet connectivity, and resource allocation (UNESCO, 2025). Without accurate infrastructure data, governments face challenges in prioritizing school development projects.
- 4. **Teacher Performance Data:** The absence of centralized data on teacher qualifications and training leads to inconsistencies in recruitment and development programs (Zizi Afrique Foundation, 2024). Limited performance evaluation metrics further exacerbate inefficiencies in teacher deployment.

**African Case Study:** South Africa's Education Management Information System (EMIS) has helped improve data collection and analysis for school planning (Baghdady & Zaki, 2019).

#### 2.3 Agriculture Sector

Reliable agricultural data is essential for food security and climate resilience, yet many African nations lack sufficient data on crop yields, weather patterns, and farmer demographics (Makombe et al., 2023). Poor data collection methodologies contribute to inefficiencies in food production and distribution.

#### **Key Agriculture Data Gaps**

- 1. **Crop Yield & Land Use Data:** Fragmented agricultural surveys prevent accurate productivity assessments (FAO, 2023). Smallholder farmers, who form the bulk of Africa's agricultural sector, often remain undocumented in official databases.
- 2. **Climate and Weather Data:** Limited meteorological data hinders farmers' ability to prepare for climate variability (FEWS NET, 2024). Inadequate weather prediction capabilities result in unpredictable farming cycles.
- 3. **Market Access and Infrastructure:** Poor road and storage data affect agricultural supply chains (World Bank, 2024). Inefficient market linkages further reduce farmers' earnings and increase food waste.

**African Case Study:** Kenya's satellite-based agricultural monitoring has enhanced food security planning (FAO, 2025).

#### 2.4 Infrastructure Sector

Infrastructure data is crucial for economic development, yet gaps in transportation, energy, and water access limit Africa's growth potential (Mo Ibrahim Foundation, 2023).

#### **Key Infrastructure Data Gaps**

- 1. **Energy Distribution:** Poorly documented electrification data hinders expansion efforts (Chinzara, 2023). The lack of grid connection records prevents efficient energy investment planning.
- 2. **Urban Development & Transport:** Lack of real-time mapping affects road and public transit planning (Mo Ibrahim Foundation, 2023).
- 3. **Water and Sanitation Data:** Incomplete records impede progress toward SDG targets (Chinzara, 2023).

**African Case Study:** Ghana's GIS-based urban planning system has improved land use tracking (Mooney & Corcoran, 2018).

# 3.Learning from Advanced Countries

Many developed nations have successfully addressed data gaps in key sectors by implementing innovative strategies, robust data governance policies, and advanced technological solutions. These approaches provide valuable lessons for African countries striving to bridge their data gaps. The approaches used by advanced countries are discussed below:

#### 3.1 Healthcare Sector

Countries such as the United States, the United Kingdom, and Germany have invested heavily in **Electronic Health Records (EHRs)** and **interoperable health information systems** to improve patient care and resource allocation (Greenhalgh et al., 2019). The United States' **HITECH Act** promoted the widespread adoption of digital health records, enhancing data sharing across hospitals and clinics (Blumenthal & Tavenner, 2010). Similarly, AI-powered diagnostics in China have improved disease detection and treatment outcomes (Esteva et al., 2019). **Lessons for Africa** include the need for national EHR policies and AI-driven healthcare analytics.

#### 3.2 Agriculture Sector

Countries like the Netherlands and the United States have leveraged precision agriculture technologies, including remote sensing, IoT-enabled devices, and blockchain-based supply chains (Gebbers & Adamchuk, 2010). The European Union's Common Agricultural Policy (CAP) has structured data collection on farm productivity, enabling better resource distribution (European Commission, 2020). Lessons for Africa include adopting smart farming solutions and integrating digital platforms for crop monitoring and market access.

#### 3.3 Education Sector

Finland and Singapore have successfully implemented **Education Management Information Systems (EMIS)** to track student progress, teacher performance, and school infrastructure development (Sahlberg, 2019). AI-driven learning analytics have enabled personalized education plans in the United States and the UK (Luan et al., 2020). **Lessons for Africa** include **developing national education databases** and using **AI-powered analytics** to enhance student learning outcomes.

#### 3.4 Infrastructure Sector

Japan and Germany have pioneered **smart infrastructure systems** that use real-time data analytics for transportation, energy distribution, and urban planning (Shibata & Kodama, 2020). Germany's **Energiewende policy** has integrated smart grids with real-time electricity consumption data, reducing energy waste (Morris & Pehnt, 2016). **Lessons for Africa** include investing in **geospatial mapping** and **IoT-based infrastructure monitoring** to enhance efficiency and sustainability.

# 4. Strategies to Bridge Data Gaps in Africa

Addressing Africa's data challenges requires a **multifaceted approach**, integrating policy reforms, technology adoption, and collaborative efforts between stakeholders. The strategies to bridge these gaps are explained below:

### 4.1 Strengthening Policy and Governance Frameworks

Governments must establish **clear data governance policies** that promote **standardized data collection, reporting, and open access** (African Union, 2021). National data policies, such as those in **Kenya and South Africa**, offer models for broader **continental adoption** (Makanga et al., 2021).

### 4.2 Leveraging Technological Innovations

Emerging technologies like **Artificial Intelligence** (**AI**), **Big Data Analytics**, **Blockchain**, and **IoT** can **revolutionize data collection and management**. Rwanda's use of AI in **health monitoring** and satellite-based agricultural tracking are examples of how innovation can improve data accuracy (Twahirwa Rutikanga, 2022).

### **4.3 Expanding Public-Private Partnerships (PPPs)**

Collaborations between **governments**, NGOs, and private entities can facilitate data-sharing **efforts**. Initiatives like **The Global Health Data Exchange (GHDx)** and **The Demographic and Health Surveys (DHS) Program** have improved data availability in the health sector (IHME, 2020). Expanding such partnerships is crucial for Africa.

# **4.4 Community-Driven Data Collection Approaches**

Local communities play a vital role in **grassroots-level data collection**. Crowdsourced initiatives like **OpenStreetMap (OSM)** and **citizen science projects** have successfully

improved geospatial data accuracy in African urban centers (Mooney & Corcoran, 2018). Encouraging community participation can enhance data inclusivity.

### 4.5 Investing in Capacity Building and Data Literacy

Strengthening statistical agencies and training data professionals is crucial. Organizations such as the African Centre for Statistics (ACS) and UNESCO Institute for Statistics (UIS) have supported data literacy programs across Africa (UNESCO, 2021). Governments must prioritize such initiatives to build sustainable data ecosystems.

# 5. Methodology

### **5.1 Research Questions**

To guide the analysis of data gaps across different sectors, the following research questions were formulated:

- 1. What are the most significant data gaps in each domain?
- 2. How do data availability and quality impact decision-making in various domains?
- 3. What are the primary challenges in accessing and using data in each domain?
- 4. How can data standards and interoperability be improved across domains?
- 5. What are the implications of data gaps on policy-making, business operations, and societal outcomes?
- 6. Which data sources are most commonly used across domains?
- 7. What is the geographic distribution of datasets?
- 8. How do data gaps impact interoperability?

# **5.2 Data Collection Approach**

To address the above research questions, a structured data collection approach was adopted. This involved identifying African datasets with missing or incomplete information across the key

sectors of **healthcare**, **education**, **agriculture**, **and infrastructure**. Details were then filled in using a set of predefined variables to ensure a comprehensive and standardized dataset.

The variables used for this process included:

- **Domain:** The specific sector of focus (e.g., healthcare, agriculture, education).
- **Sub-Domain:** A more specific category within the domain.
- **Data Type:** Structured, unstructured, or semi-structured.
- **Data Source:** The origin of the data (e.g., government database, survey, API).
- Data Availability: Whether the data is abundant, limited, or non-existent.
- **Data Quality:** The accuracy, completeness, and consistency of the data.
- **Data Format:** How the data is stored (e.g., CSV, JSON, XML).
- Access Constraints: Any restrictions on accessing the data.
- **Update Frequency:** How often the data is updated (e.g., real-time, monthly, annually).
- **Data Standard:** The common frameworks or specifications for ensuring data consistency.
- **Gap Analysis:** Identifying specific limitations in the dataset.
- **Geographic Scope:** The level of coverage (e.g., regional, national, global).
- **Temporal Scope:** The time period covered (e.g., historical, real-time, predictive).
- **Data Governance:** Policies in place to manage and oversee the data.
- **Interoperability:** The ability of the data to integrate seamlessly across different systems.

### **5.3 Proposed Data Collection Tool**

The **primary tool for data collection and storage** was Microsoft Excel. This tool was chosen due to its **flexibility**, **ease of use**, **and ability to handle structured data** efficiently. Each dataset was compiled into an Excel sheet, with columns corresponding to the above-listed variables. Excel's built-in features for sorting, filtering, and data validation were utilized to ensure **consistency and accuracy**.

#### **5.4 Proposed Analytics Tool**

For **analyzing** the collected data, Python-based data analytics tools such as **Pandas and NumPy** were proposed. These tools allow for:

- Data cleaning and preprocessing to ensure completeness and consistency.
- Exploratory Data Analysis (EDA) to identify trends and patterns within the datasets.
- Gap identification and visualization, using libraries like Matplotlib and Seaborn to highlight missing information and inconsistencies.

# 6. Summary of Key Research Gaps

### 6.1 Gaps in Data Standardization and Real-Time Availability

One of the most pressing research gaps is the lack of standardized and real-time data collection methodologies across key sectors. Many African countries lack cohesive national data policies, leading to inconsistencies in reporting and measurement frameworks (Mo Ibrahim Foundation, 2023). This makes cross-country comparisons difficult and affects data-driven decision-making. Healthcare, agriculture, education, and infrastructure sectors all suffer from delayed, incomplete, and unreliable data reporting, which hinders timely intervention planning.

# 6.2 Lack of African-Specific Case Studies

Most of the literature on data governance and usage comes from global institutions, with limited **Africa-specific studies** (World Bank, 2024). Many existing frameworks for data-driven decision-making are derived from **Western economies**, which do not always align with the unique **economic**, **social**, **and political contexts of African nations**. Expanding research on **successful data initiatives within Africa**, such as Kenya's satellite-driven agricultural monitoring or Rwanda's AI-powered health systems, is necessary for tailored policy recommendations.

# **6.3** Weak Coordination Among Stakeholders

A critical gap in Africa's data ecosystem is the **lack of coordination between governments**, **private entities**, **research institutions**, **and international organizations** (African Union, 2021). **Fragmented data ownership** prevents efficient data-sharing, leading to duplicated efforts and

inefficient resource allocation. There is an urgent need for **stronger governance frameworks** that **enable seamless data collaboration** between different stakeholders.

### **6.4 Limited Use of Advanced Technologies**

Despite global advancements in **AI**, **blockchain**, **and IoT**, many African countries are **slow to integrate these technologies into data collection and management** (Twahirwa Rutikanga, 2022). Governments and organizations often rely on **traditional**, **manual data collection methods**, which are **costly**, **time-consuming**, **and prone to errors**. Research is needed to explore how **emerging technologies can be adapted** to fit Africa's resource constraints and infrastructure challenges.

### 6.5 The Need for Strengthening Public-Private Partnerships (PPPs)

Public-private collaborations remain underutilized in Africa's data ecosystem (IHME, 2020). While many successful data initiatives in healthcare and agriculture have been led by international organizations or private tech firms, government engagement remains low. Strengthening PPPs could significantly improve data quality, accessibility, and sustainability, especially in underfunded sectors like education and infrastructure.

### 6.6 Addressing Digital Literacy and Capacity Building

A major challenge in Africa's data landscape is the **shortage of skilled professionals in data science**, **analytics**, **and governance** (UNESCO, 2021). Many national statistics offices and research institutions lack the expertise needed to **effectively collect**, **analyze**, **and interpret complex datasets**. Investment in **capacity-building programs**, **digital literacy training**, **and STEM education initiatives** will be crucial to closing this gap

# 7. Conclusion

Data gaps remain a significant challenge across Africa's key sectors, hindering **evidence-based policymaking** and **socioeconomic development**. However, insights from **global best practices** demonstrate that leveraging **AI**, **big data**, **and public-private collaborations** can significantly improve data collection and utilization. Governments, private entities, and civil society must work together to implement **comprehensive data governance frameworks**, **integrate digital technologies**, and **invest in capacity-building initiatives**. By prioritizing these actions, Africa can bridge its data gaps and unlock its full development potential.

### 8. REFERENCE LIST

- Adjadeh, R. (2024). AD768: Assessing the gaps: Africans look for greater progress on education. Afrobarometer. Available at: <a href="https://www.afrobarometer.org/publication/ad768-assessing-the-gaps-africans-look-for-greater-progress-on-education/">https://www.afrobarometer.org/publication/ad768-assessing-the-gaps-africans-look-for-greater-progress-on-education/</a>
- 2. Afrobarometer. (2024). Assessing the gaps: Africans look for greater progress on education. Afrobarometer Dispatch No. 768. Retrieved from <a href="https://www.afrobarometer.org/wp-content/uploads/2024/02/AD768-PAP9-Assessing-gaps-Africans-want-progress-on-education-Afrobarometer-9feb24.pdf">https://www.afrobarometer.org/wp-content/uploads/2024/02/AD768-PAP9-Assessing-gaps-Africans-want-progress-on-education-Afrobarometer-9feb24.pdf</a>
- 3. Baghdady, A., & Zaki, O. (2019). Secondary education governance in Sub-Saharan Africa. Mastercard Foundation. Available at: <a href="https://mastercardfdn.org/wp-content/uploads/2019/03/SEA-Education-Governance-Final\_Feb-28.pdf">https://mastercardfdn.org/wp-content/uploads/2019/03/SEA-Education-Governance-Final\_Feb-28.pdf</a>
- 4. Carolan, M. (2020). Digitization, transparency, and traceability in agri-food supply chains. Journal of Rural Studies, 78, 304-312.
- Chinzara, Z., (2023). Infrastructure in Africa: How institutional reforms can attract more
  private investment. International Finance Corporation. Retrieved
  from <a href="https://www.ifc.org/content/dam/ifc/doc/2023/working-paper-infrastructure-in-africa.pdf">https://www.ifc.org/content/dam/ifc/doc/2023/working-paper-infrastructure-in-africa.pdf</a>
- 6. CIPESA. (2023). How technology is impacting health data governance in Africa: The case of Uganda. Available at: <a href="https://cipesa.org/2023/07/how-technology-is-impacting-health-data-governance-in-africa-the-case-of-uganda/">https://cipesa.org/2023/07/how-technology-is-impacting-health-data-governance-in-africa-the-case-of-uganda/</a>
- 7. CSM Tech. (2024). The healthcare data crisis in Africa and how to mitigate it. Available at: <a href="https://www.csm.tech/blog-details/the-healthcare-data-crisis-in-africa-and-how-to-mitigate-it">https://www.csm.tech/blog-details/the-healthcare-data-crisis-in-africa-and-how-to-mitigate-it</a>
- 8. Davis, S., & Roberts, H. (2022). Data gaps in African educational systems. International Journal of Education, 50(2), 62-77.
- 9. Dell, J. (2023). What we mean by data gaps. LinkedIn. Available at: <a href="https://www.linkedin.com/pulse/what-we-mean-data-gaps-jon-dell-mba/">https://www.linkedin.com/pulse/what-we-mean-data-gaps-jon-dell-mba/</a>
- 10. Development Gateway: An IREX Venture. (2024). Harnessing data to enhance agricultural & food security. Retrieved from https://developmentgateway.org/expertise/agriculture/
- 11. Development Gateway: An IREX Venture. (2024). Harnessing data to enhance agricultural & food security. Retrieved from https://developmentgateway.org/expertise/agriculture/

- 12. Esteva, A., Kuprel, B., Novoa, R. A., Ko, J., Swetter, S. M., Blau, H. M., & Thrun, S. (2019). Dermatologist-level classification of skin cancer with deep neural networks. Nature, 542(7639), 115-118.
- 13. European Commission. (2020). The Common Agricultural Policy (CAP) and its role in EU food security.
- 14. Famine Early Warning Systems Network (FEWS NET). (2024, October 7). Bridging Africa's agricultural data gap. Medium. Retrieved from <a href="https://medium.com/@fewsnet/bridging-africas-agricultural-data-gap-a88a8e2355d7">https://medium.com/@fewsnet/bridging-africas-agricultural-data-gap-a88a8e2355d7</a>
- 15. Food and Agriculture Organization (FAO). (2023). A lack of basic agricultural data is holding African countries back. Retrieved from <a href="https://www.fao.org/family-farming/detail/en/c/1412330/">https://www.fao.org/family-farming/detail/en/c/1412330/</a>
- 16. Food and Agriculture Organization (FAO). (2025). FAO supports Zimbabwe to prepare for the National Agricultural and Livestock Census. Retrieved from <a href="https://www.fao.org/africa/news-stories/news-detail/fao-supports-zimbabwe-to-prepare-for-the-national-agricultural-and-livestock-census/en">https://www.fao.org/africa/news-stories/news-detail/fao-supports-zimbabwe-to-prepare-for-the-national-agricultural-and-livestock-census/en</a>
- 17. Gebbers, R., & Adamchuk, V. I. (2010). Precision agriculture and food security. Science, 327(5967), 828-831.
- 18. Greenhalgh, T., Vijayaraghavan, S., Wherton, J., Shaw, S., Byrne, E., Campbell-Richards, D., & Hodkinson, I. (2019). Interoperability and data sharing in the NHS: Challenges and solutions. BMJ, 367, 15314.
- 19. Healthcare Financial Management Association. (2021). Unleashing the power of healthcare workforce data. HFMA. Available at: <a href="https://www.hfma.org/finance-and-business-strategy/analytics/unleashing-the-power-of-healthcare-workforce-data/">https://www.hfma.org/finance-and-business-strategy/analytics/unleashing-the-power-of-healthcare-workforce-data/</a>
- 20. Ismaila, L. (2023). State of medical data in Africa: A case of barriers to science in Africa. The DataSphere. Available at: <a href="https://www.thedatasphere.org/news/state-of-medical-data-in-africa-a-case-of-barriers-to-science-in-africa/">https://www.thedatasphere.org/news/state-of-medical-data-in-africa-a-case-of-barriers-to-science-in-africa/</a>
- 21. Jiang, F., Jiang, Y., Zhi, H., Dong, Y., Li, H., Ma, S., ... & Wang, Y. (2017). Artificial intelligence in healthcare: Past, present and future. Stroke and Vascular Neurology, 2(4), 230-243.
- 22. Kabeta, S. H., Chala, T. K., & Tafese, F. (2023). Medical equipment management in general hospitals: Experience of Tulu Bolo General Hospital, South West Shoa Zone, Central Ethiopia. Medical Devices (Auckland, N.Z.), 16, 57–70. doi: 10.2147/MDER.S398933. Available at: https://pmc.ncbi.nlm.nih.gov/articles/PMC10029930/

- 23. Kamilaris, A., Kartakoullis, A., & Prenafeta-Boldú, F. X. (2019). A review on the practice of big data analysis in agriculture. Computers and Electronics in Agriculture, 163, 104–121.
- 24. Kirigia, J. M., & Barry, S. (2008). Health challenges in Africa and the way forward. ResearchGate. Available at: <a href="https://www.researchgate.net/publication/23675192">https://www.researchgate.net/publication/23675192</a> Health Challenges in Africa and the way forward
- 25. Larkin, C., et al. (2018). Access to higher education in Africa: Challenges and opportunities. Journal of Higher Education in Africa, 16(1), 21-38.
- 26. Luan, H., Tsai, C. C., & Chang, H. Y. (2020). The impact of learning analytics on student learning performance: A meta-analysis of empirical studies. Educational Technology & Society, 23(2), 125-142.
- 27. Matchaya, G., Makombe, T., & Mihaylova, N. G. (2023). Data challenges and opportunities for food systems transformation in Africa. In ReSAKSS Annual Trends and Outlook Report. Retrieved from <a href="https://www.resakss.org/sites/default/files/2023\_ator\_individual\_chapters/Chapter%2011\_ReSAKSS\_AW\_ATOR\_2023.pdf">https://www.resakss.org/sites/default/files/2023\_ator\_individual\_chapters/Chapter%2011\_ReSAKSS\_AW\_ATOR\_2023.pdf</a>
- 28. McLoughlin, K., & Dwolatzky, B. (2014). The information gap in higher education in South Africa. South African Journal of Higher Education, 28(2), 584–604. Available at: <a href="https://journals.co.za/doi/epdf/10.10520/EJC153539">https://journals.co.za/doi/epdf/10.10520/EJC153539</a>
- 29. Miller, H. J., & Goodchild, M. F. (2015). Data-driven geography. GeoJournal, 80(4), 449-461.
- 30. Morris, C., & Pehnt, M. (2016). Energy transition: The German Energiewende. Heinrich Böll Stiftung.
- 31. Mo Ibrahim Foundation (2024). 2023 IIAG Series Report: Closing Data Gaps to Accelerate Africa's Transformation. Available at: <a href="https://mo.ibrahim.foundation/sites/default/files/2024-01/2023-iiag-series-report.pdf">https://mo.ibrahim.foundation/sites/default/files/2024-01/2023-iiag-series-report.pdf</a>
- 32. Mukasa, A. N., Woldemichael, A. D., Salami, A. O., & Simpasa, A. M. (2017). Africa's agricultural transformation: Identifying priority areas and overcoming challenges. African Development Bank.
- 33. Muniu, S. (2024). Common challenges in African health programs: Utilization, data management, and supply chain issues. Aidspan. Available at: <a href="https://aidspan.org/common-challenges-in-african-health-programs-utilization-data-management-and-supply-chain-issues">https://aidspan.org/common-challenges-in-african-health-programs-utilization-data-management-and-supply-chain-issues</a>
- 34. Musa, S.M., Haruna, U.A., Manirambona, E., Eshun, G., Ahmad, D.M., Dada, D.A., Gololo, A.A., Musa, S.S., Abdulkadir, A.K., and Lucero-Prisno, D.E. (2023). Paucity of Health Data in Africa: An Obstacle to Digital Health Implementation and Evidence-

- Based Practice. Public Health Reviews, 44, 1605821. Available at: <a href="https://pmc.ncbi.nlm.nih.gov/articles/PMC10495562/">https://pmc.ncbi.nlm.nih.gov/articles/PMC10495562/</a>
- 35. Ng, W. (2017). Emerging technologies for learning: Innovation in the 21st-century classroom. Springer Science & Business Media.
- 36. Oleribe, O. O., et al. (2019). Identifying key challenges facing healthcare systems in Africa and potential solutions. PMC. Available at: <a href="https://pmc.ncbi.nlm.nih.gov/articles/PMC6844097/">https://pmc.ncbi.nlm.nih.gov/articles/PMC6844097/</a>
- 37. Sahlberg, P. (2015). Finnish lessons 2.0: What can the world learn from educational change in Finland? Teachers College Press.
- 38. Shibata, S., & Kodama, K. (2020). Smart cities and urban infrastructure development in Japan: An overview. International Journal of Urban Planning, 34(3), 245-259.
- 39. UNESCO Institute of Statistics. (2025). Education in Africa. Available at: <a href="https://uis.unesco.org/en/topic/education-africa">https://uis.unesco.org/en/topic/education-africa</a>
- 40. Westerink, H. J., Bresser, C. C., Garvelink, M. M., van Uden-Kraan, C. F., Zouitni, O., Bart, H. A. J., van der Wees, P. J., & van der Nat, P. B. (2023). The use of outcome data in patient consultations from the healthcare professionals' and patients' perspectives: A mixed methods study. Journal of Evaluation in Clinical Practice. Available at: <a href="https://www.sciencedirect.com/science/article/pii/S073839912300424X">https://www.sciencedirect.com/science/article/pii/S073839912300424X</a>
- 41. World Economic Forum. (2024, February 12). How UNESCO is trying to plug the data gap in global education. Retrieved from https://www.weforum.org/stories/2024/02/unesco-data-gap-global-education/
- 42. World Health Organization (2019). Countries to Strengthen Health Data collection, analysis and use in support of resilient health systems. Available at: <a href="https://www.afro.who.int/news/countries-strengthen-health-data-collection-analysis-and-use-support-resilient-health-systems">https://www.afro.who.int/news/countries-strengthen-health-data-collection-analysis-and-use-support-resilient-health-systems</a>
- 43. Zhu, L., et al. (2020). Data gaps in teacher performance in sub-Saharan Africa. Journal of Educational Research, 41(2), 123-136.