

Certificate I: Understanding AI and Machine Learning in Africa

Course AIMLO1: Artificial Intelligence – Past, Present, and Future

Module 4: Future Challenges

Lecture 3: Social and Ethical Aspects of AI

Carnegie Mellon University
Africa

Learning Objectives

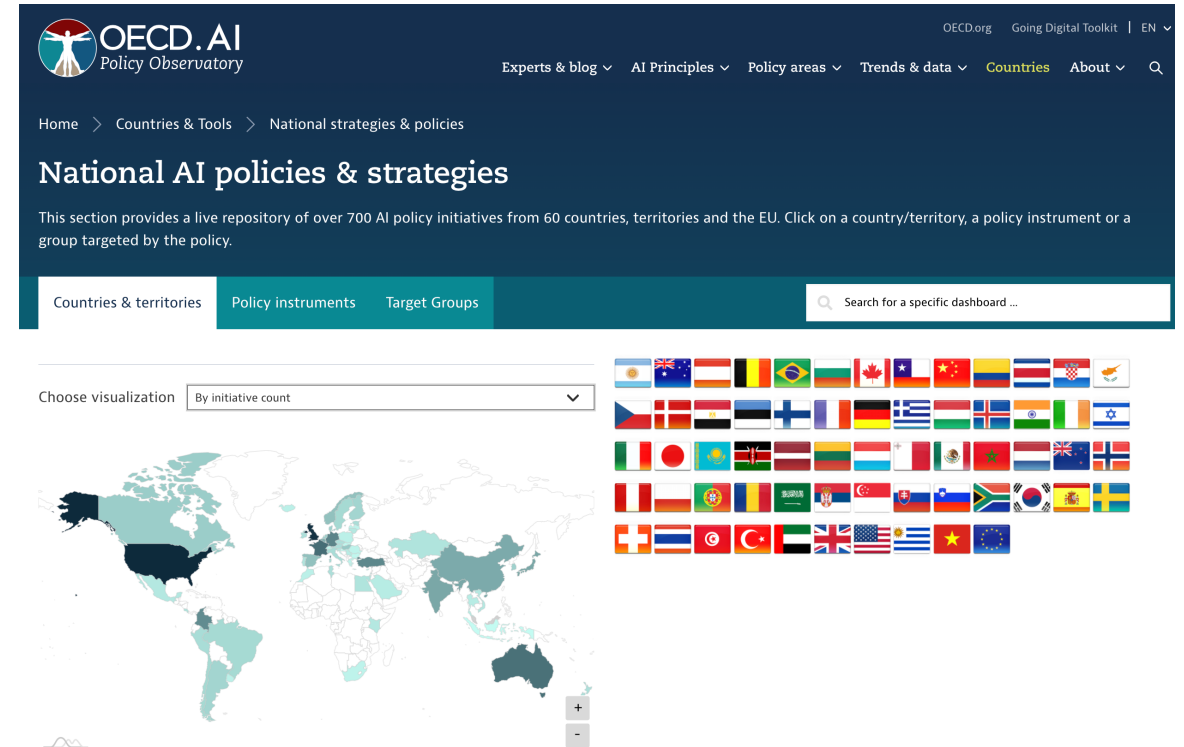
1. Highlight and illustrate the relevance of AI to Africa
2. Identify the ways in which AI can have a negative impact on developing countries
3. Discuss ethical concerns, such as the dangers of fake misinformation, the problem of gender and racial bias, and the imperative of democratization in AI

Lecture Contents

1. The positive impact of AI in Africa
2. Premature deindustrialization
3. Ethical imperatives
4. Democratization in AI
5. Lecture summary
6. Recommended reading & references

The Positive Impact of AI in Africa

- AI can benefit everyone
- The examples so far focused on applications in the developed world
- Most of the national strategies on AI have been created by governments in developed countries
- AI is just as relevant to Africa



<https://oecd.ai/en/dashboards>

AI in Africa for Sustainable Economic Development

2020 ACM International Conference on Artificial Intelligence in Finance (ICAIF) Workshop

14th of October 2020 (8am -12:30pm ET)

Artificial intelligence (AI), facilitated by easier data collection and improved computing resources, is shaping the dynamics of many sectors that are closely linked with achieving the Sustainable Development Goals. Many African countries have tremendous opportunities to use AI in a number of key sectors including finance, agriculture, health, infrastructure and food security. However, the lack of expertise and capacity, as well as impacts of the current Covid19 pandemic, pose significant challenges. Despite the extensive promises of AI to transform economies and expedite development, the challenges and adverse impacts need to be studied thoroughly.

The Positive Impact of AI in Africa

AI is having an increasingly positive impact in Africa in sectors such as

See (Novitske, 2018; Access Partnership, 2018; Delmus Alupo et al., in press)

- Energy
- Healthcare
- Agriculture
- Public services
- Financial services



<https://www.mamopanel.org/resources/reports-and-briefings/byte-byte-policy-innovation-transforming-africas-f/>

The Positive Impact of AI in Africa

It has the potential to

- Drive economic growth, development, and democratization,
- Reduce poverty
- Improve education
- Support healthcare delivery
- Increase food production
- Improve the capacity of existing road infrastructure by increasing traffic flow
- Improve public services, and
- Improve the quality of life of people with disabilities

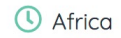
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LET'S TALK

AGRI-INDUSTRY PROCESS AUTOMATION

We deliver detection capabilities beyond the scope of human noses and eyes as well as analysis that drives inputs to the manufacturing processes. For the first time in tea history, we are leveraging the Industrial Internet of Things (IoT) to improve quality, control, and revenues in the African tea value-chain.

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DRONES ON THE HORIZON

TRANSFORMING AFRICA'S AGRICULTURE





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IMPACT

Using Satellite Images and Artificial Intelligence to Improve Agricultural Resilience

<https://www.rti.org/impact/using-satellite-images-and-artificial-intelligence-improve-agricultural-resilience>

Market Watch

May 8, 2019

Kenya: Microsoft Brings FarmBeats, AI, Edge Computing, Drones to Africa

By [Microsoft Care GH](#)

Share



Microsoft is expanding the pilot program of FarmBeats to Africa, starting in Kenya, [reports Microsoft Cares GH](#). The program in Nairobi will reportedly be focused on addressing the specific challenges of farming in Africa with the intent of expanding to other African countries.

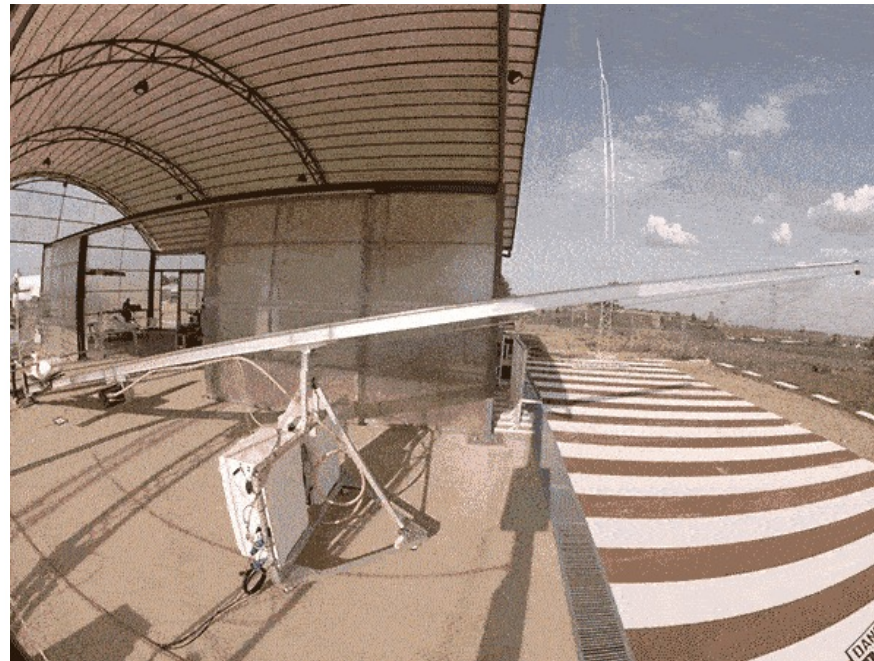
In the Air With Zipline's Medical Delivery Drones

Commercial operations in Rwanda prove the company can deliver emergency blood packs in minutes, rather than hours

By Evan Ackerman and Michael Koziol

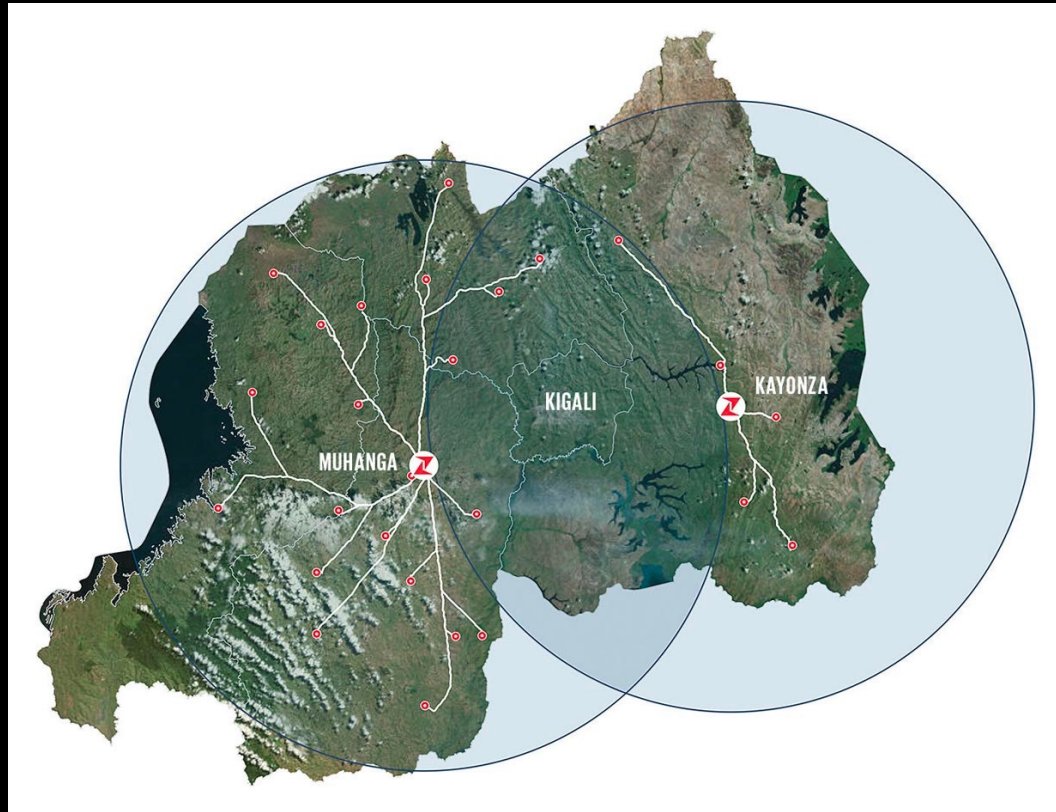
TECH EXPEDITION

**East Africa's
Big Bet
On Drones**



Gif: IEEE Spectrum

Zipline Medical Delivery Drones

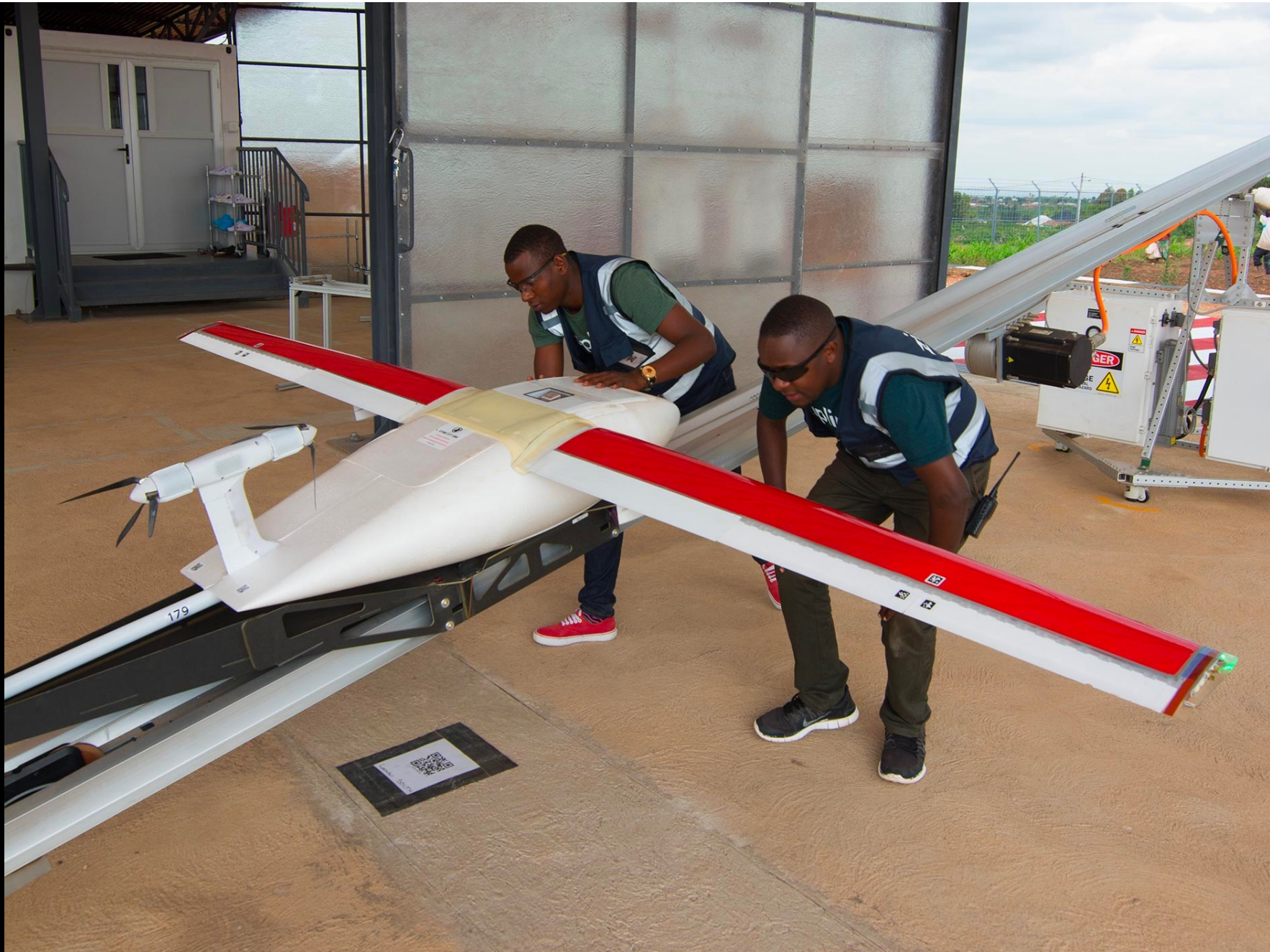


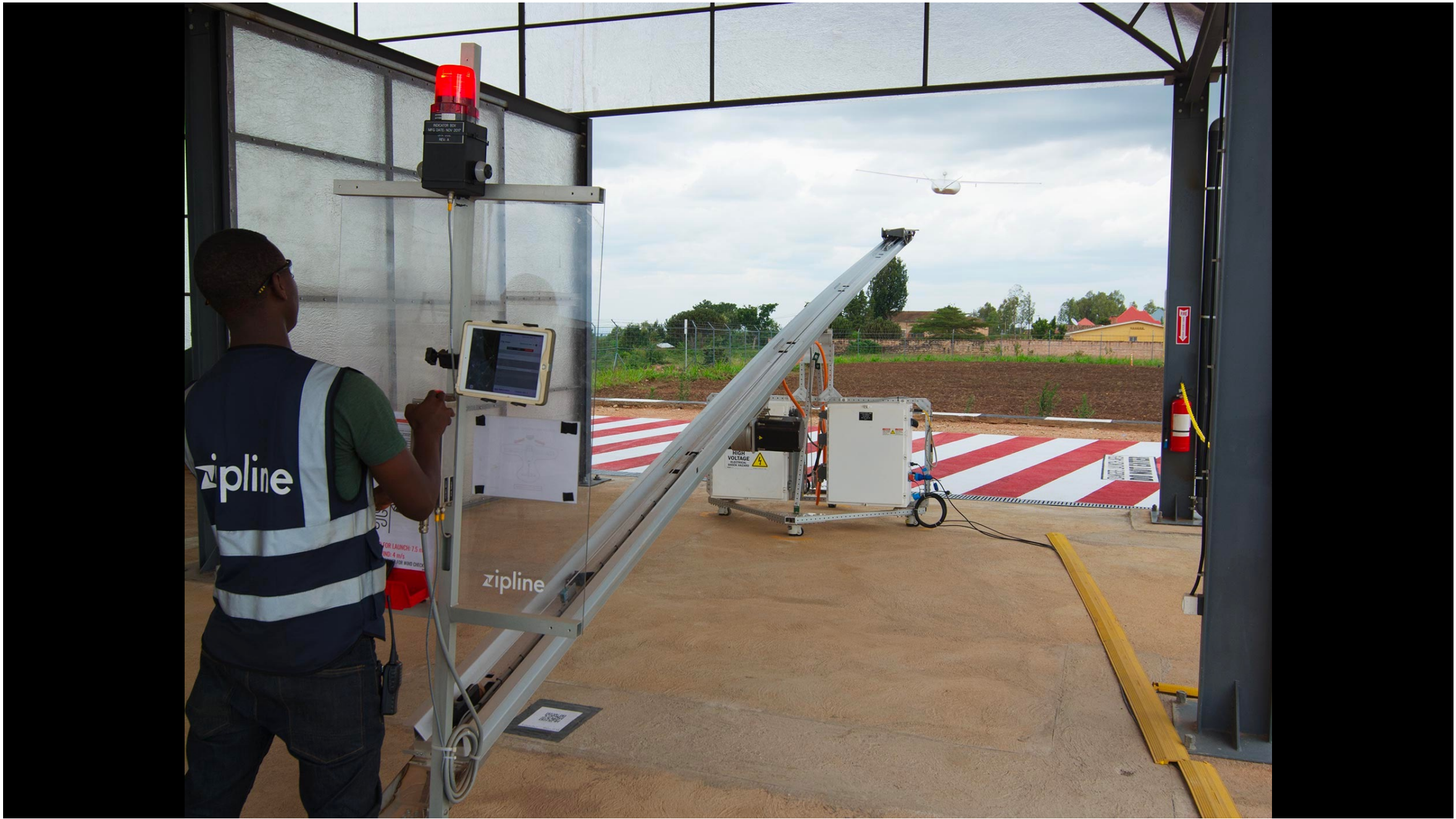












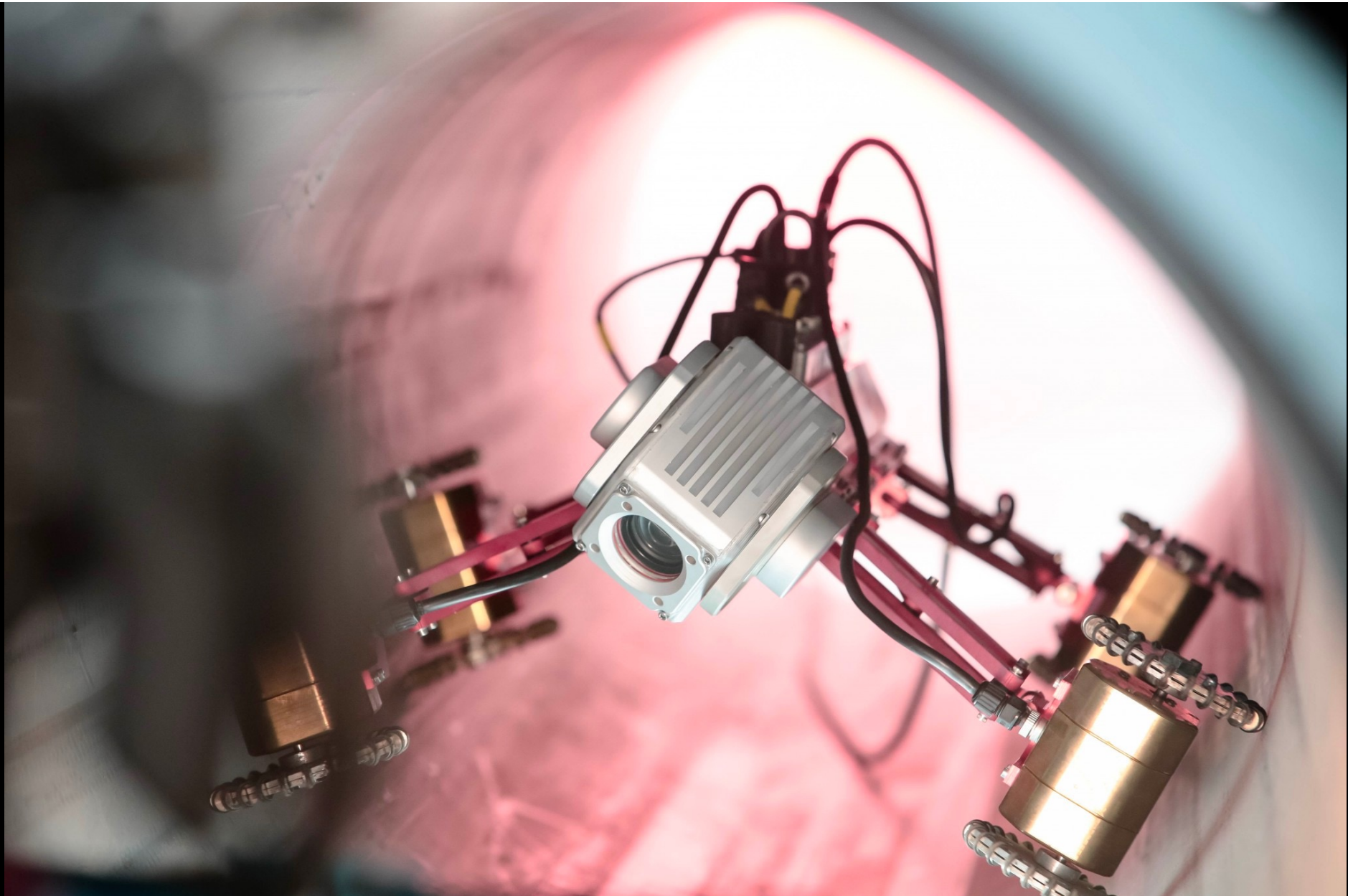




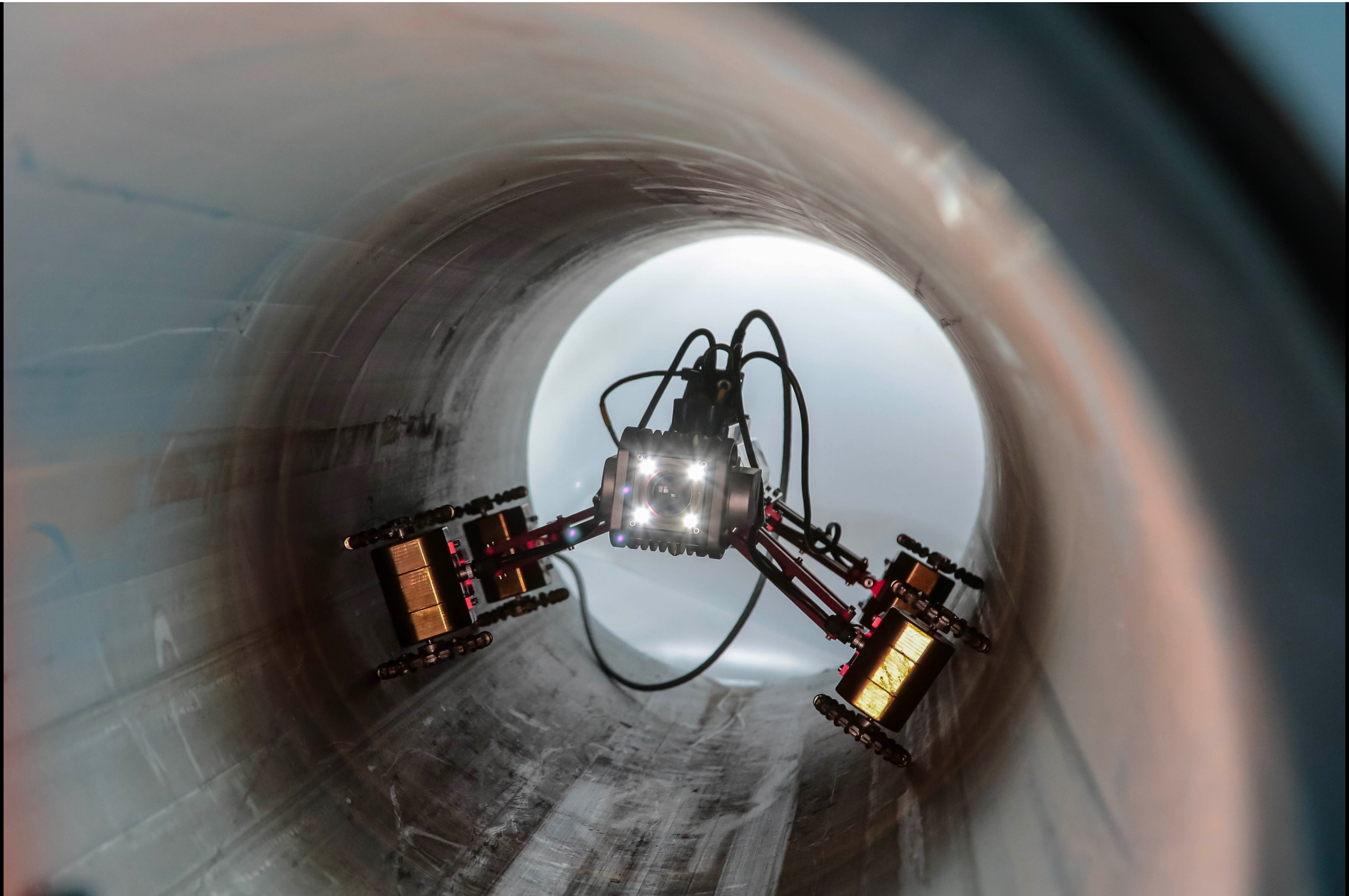








<https://www.ryonic.io/products/pipeline-inspection-crawlers/rmis-m8/>



<https://www.ryonic.io/products/pipeline-inspection-crawlers/rmis-m8/>

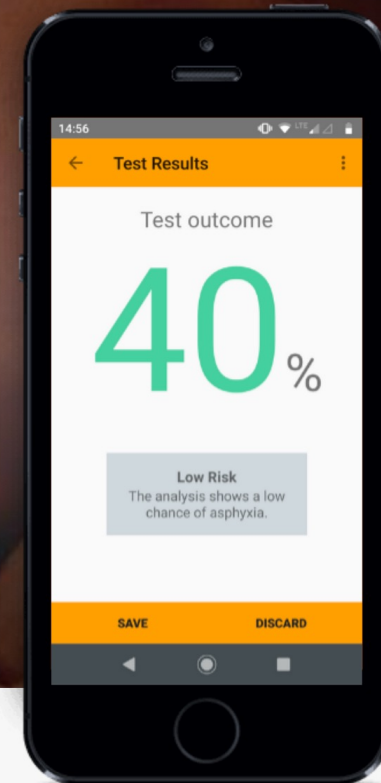


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GIVING HOPE TO NEWBORNS

Ubenwa analyses a baby's cry to detect early signs of anomalies which could be fatal such as asphyxia or brain-injury.

[Stay in touch](#)



WHAT IS UBENWA?

Ubenwa is a mobile app that analyses the cry sounds of a newborn to detect early signs of perinatal asphyxia - a **leading cause of neonatal disability and death**. The app uses machine learning to identify the changing acoustic patterns in the cries of newborns who are at risk of brain damage due to asphyxia. It alerts care-givers of infants at risk, allowing them to apply necessary treatment and/or make an early referral to tertiary care facilities. See our short video below to learn more.



Charles Onu
Founder and AI Research lead



Photos from the Ubenwa clinical study in Nigeria. Click [here](#) for more photos.



ABOUT

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CAREERS

LOGIN

PRIVACY



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GET APP

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ULIMA IS A MOBILE PLATFORM DESIGNED SPECIFICALLY FOR FARMERS, AGRO-DEALERS,
AND THE BROADER AGRICULTURAL COMMUNITY.

<http://ulima.co/>



We aim to **secure** Africa's food production

How we do it



COVID-19 Mifugo Xtension



Our Impact





We celebrate small holder farmers!

How we do it



COVID-19 Mifugo Xtension



Our Impact





How we do it



COVID-19 Mifugo Xtension




Our Impact





Robotics


With our Robotics programs, your child will enjoy exciting projects that use 21st century skills.




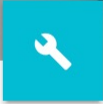
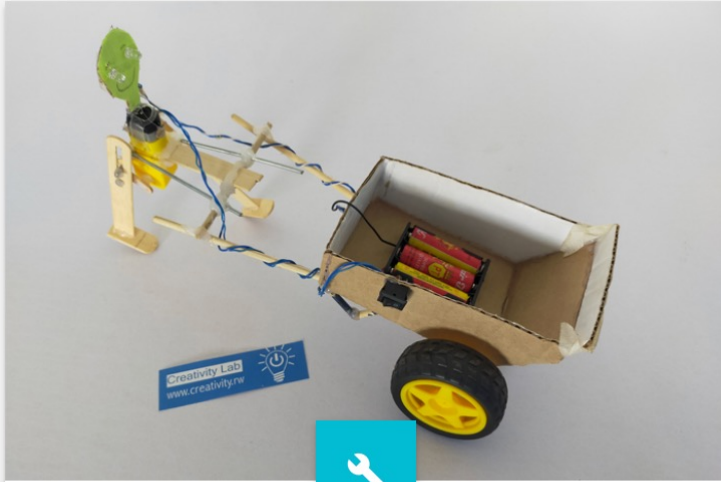
 Robotics for Kids



 Age group 7 - 13 Years old

 Language English

 Level Primary education



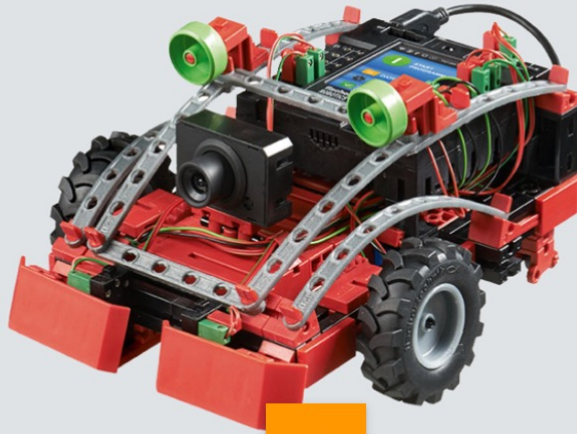
DIY basic robots

After this lesson, child will be able to:

- Define what a robot is
- Describe the main components of a robot
- Explain how engineers apply robotics to solve real-world problems
- Apply practical math, mechanical construction and critical thinking to make a simple robot

 Age 9 - 13

 Locked



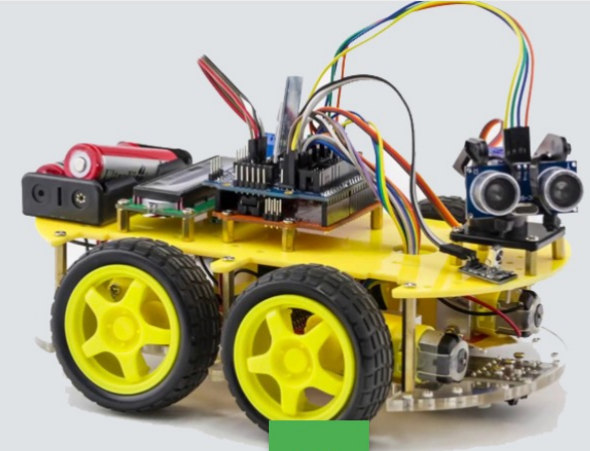
Fischertechnik TXT Advanced

After this lesson, child will be able to:

- Describe common robot applications
- Identify the different parts of robot
- Do robots assembly and programming
- Use sensors to estimate a robot's condition and environment
- Make Soccer robot, Detection robot, Mobile robot, Camera man robot,...

 Age 7 - 13

 Locked



DIY programmable robots

After this lesson, child will be able to:

- Discover different categories of robots and how they interact with their surroundings
- Illustrate a robot's form based upon a specific set of function
- Program and operate a robot to address a challenge

 Age 10 - 13

 Locked

Premature Deindustrialization

The deployment of AI in **developed countries** can have a severe **negative impact** on **developing countries** due to the phenomenon known as **premature deindustrialization** (Rodrik, 2016; UNCTAD, 2016)

Developing countries lose their competitive advantage in manufacturing due to the **lower cost automation in developed countries**

Developing countries miss out on the economic benefits that **developed countries** enjoyed as their workforces moved from low-value work to manufacturing before progressing to a post-industrial service economy

Premature Deindustrialization

Developing countries are increasingly likely **not to have the opportunity** for rapid economic growth by shifting workers from farms to factory jobs because

- Automation undermines the labor cost advantage
- Developments in robotics and additive manufacturing allow companies in advanced economies to locate production closer to domestic markets in automated factories

Ethical Imperatives

- AI can have a negative impact, intentionally or unintentionally, in many other ways
 - For example, by fomenting **religious**, **ethnic**, **social**, and **political divisions** through fake misinformation created by deep networks (Besaw and Filitz, 2019)
- Ethical imperatives
 - Eliminate **bias** of all types, including gender and racial bias
 - Foster **trust**
 - Safeguard **privacy**
 - Guarantee **confidentiality**
 - Maximize **explainability**

Gender and Racial Bias

- The data that are used to train the AI models can be **biased**
- Thereby resulting in discrimination against people on the basis of **gender** or **race**
- Examples of bias against dark-skinned people include
 - Face analysis (Buolamwini and Gebru, 2018)
 - Pedestrian detection (Wilson et al., 2019)
 - Predicting recidivism (Larson et al., 2016)
- AIMLO2 Social, Ethical, Legal, and Economic Impact of AI & ML covers these issues in more detail




The tendency of a convicted criminal to reoffend

Democratization in AI

- Open access to AI technology by developers everywhere
 - In both developing and developed countries
- Training deep neural networks requires access to large expensive computational resources
 - May be out of reach of some
- Training also requires very large data sets
 - These may not be available

Democratization in AI

- Using machine learning to make the web available in **local African languages**
 - **Progress is being inhibited** for so-called “low- resourced” languages
 - Because of the lack of sufficient training data (Nekoto et al., 2020)
- This is a problem with almost all applications of machine learning in developing countries: the **lack of data**



Languages for which few digital or computational data resources exist

Lecture Summary

1. AI has the potential to improve significantly the lives of people in Africa
2. AI can be applied in many different sectors, from agriculture, to education, to healthcare
3. The adoption of AI in developed countries can make it difficult (but not impossible) for developing countries to make the transition to an industrialized economy
4. We need to be aware of the dangers of inappropriate use of AI, for example to generate fake misinformation
5. We need to ensure that the training sets for machine learning do not have any bias, such as racial bias or gender bias

Recommended Reading

Delmus Alupo C, Omeiza D, Vernon D (in press) Realizing the potential of AI in Africa. In: Ferreira MIA (ed) Towards Trustworthy Artificial Intelligence Systems, Springer.

http://vernon.eu/publications/2023_Alupo_et_al.pdf

Vernon D (2019) "Robotics and Artificial Intelligence in Africa", IEEE Robotics & Automation Magazine, Vol. 26, No. 4, pp. 131-135, December.

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<https://doi.org/10.1162/99608f92.f06c6e61>

Besaw C, Filitz J (2019) AI & global governance: AI in Africa is a double-edged sword. Tech. Rep., United Nations University.
<https://cpr.unu.edu/ai-in-africa-is-a-doubleedged-sword.html>

Buolamwini, J., & Gebru, T. (2018). Gender shades: Intersectional accuracy disparities in commercial gender classification. In Conference on fairness, accountability and transparency (pp. 77-91). Proceedings of Machine Learning Research 81:1 – 15.
<https://proceedings.mlr.press/v81/buolamwini18a/buolamwini18a.pdf>

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<https://www.propublica.org/article/how-we-analyzed-the-compas-recidivism-algorithm>

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<https://aclanthology.org/2020.findings-emnlp.195.pdf>

Novitske L (2018) The AI invasion is coming to Africa (and it's a good thing). Stanford Social Innovation Review
https://ssir.org/articles/entry/the_ai_invasion_is_coming_to_africa_and_its_a_good_thing#

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<https://oecd.ai/dashboards>

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UNCTAD (2016) Robots and industrialization in developing countries. Tech. Rep. United Nations Conference on Trade and Development (UNCTAD).
https://unctad.org/en/PublicationsLibrary/presspb2016d6_en.pdf

Wilson B, Hoffman J, Morgenstern J (2019) Predictive inequity in object detection. arXiv preprint arXiv:1902.11097
<https://arxiv.org/pdf/1902.11097.pdf>