THE ERA OF USING BIG DATA, ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING IN CENTRAL BANKS AND FINANCIAL **INSTITUTIONS: IMPLICATIONS FOR MONETARY POLICY**

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GUEST SPEAKER PRESENTATION

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Background: Take a Trip down Memory Lane

- Irving Fisher's Quantity Theory of Money, MV=PT, in 1911
- The IS-LM model, developed by John Hicks in 1937, based on Keynesian theories
- The flow of funds, Morris Copeland in 1952, and asset demand equations
- The rational expectations revolution, by Lucas and Sargent in the 1970s
- The Econometric Society, founded in December 1930, by Irving Fisher and others with data-driven methods for monetary policy
- And now, all change: Big data, Al, and ML
- Hence, we focus on the era of using Big Data, AI, and ML to better understand:
 - how advances in BD, AI, and ML are impacting Central Banks and financial institutions;
 - the implications of these developments on monetary policy
 - to create knowledge, of great benefit to central banks in the COMESA region

Road Map of this Presentation

- What do we really know about Big Data, Artificial Intelligence (AI), and Machine Learning (ML)?
- The use of Big Data, AI, and ML in Financial Institutions: Applications, opportunities, as well as challenges and risks
- The use of Big Data, AI, and ML in Central Banks: OECD and Africa
- The key applications for monetary policy, some priorities, challenges, and mitigation of risk
- Concluding remarks, key 'takeaways', and the role of the AERC
- Additional information and references to ignite research in this area

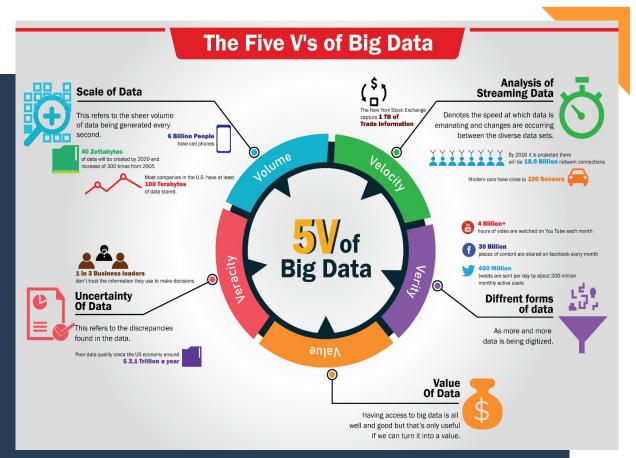


The 'Big Five' of 'Big Data'

Big data: datasets that are too large or complex to be dealt with by traditional data processing application software

Big data has 5 distinct characteristics:

- Volume scale of data; i.e., the volume of data being generated
- **Veracity** the uncertainty of data, i.e., the discrepancies inherent in the data
- **Value** it is critical not only to have access to data but also to have data that is usable
- **Velocity** the speed at which data is emanating and changing between different datasets
- Variety –different forms of data



Source: https://medium.com/@varunmishrait76/what-is-big-data-78a934cf4e4b



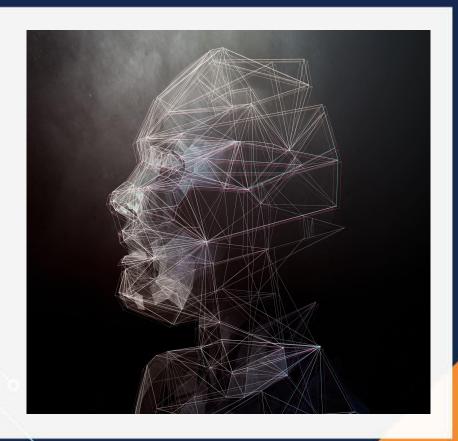


- International Finance Corporation (IFC) defines Artificial intelligence (AI) as "the science of making machines act in rational, intelligent ways"
- These smart machines simulate human intelligence processes to perform tasks that require human intelligence - tasks such as reasoning, learning, problem-solving, understanding natural language, perception, and decision-making
- Existing literature distinguishes between two types of AI:
 - Narrow AI: All systems that are designed to perform a specific task, such as image recognition or language translation
 - **General AI:** a theoretical AI that would be capable of performing any intellectual task that a human can do



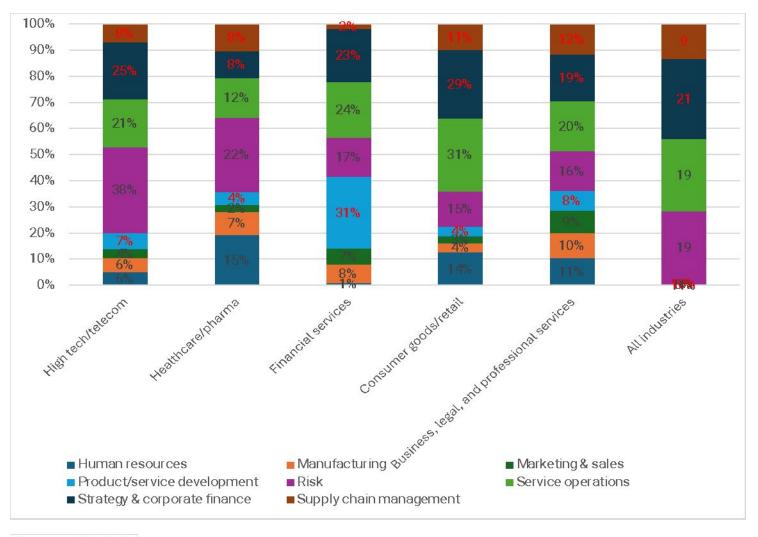
What is AI? Cont'd...

- Examples of AI are Face ID, the search algorithm, the recommendation algorithm, algorithmic trading in financial markets
- Examples of generative AI are Copilot, Google's Bard, ChatGPT, or DALL-E from OpenAI
- Their applications include but are not limited to, healthcare and pharmaceutical services, manufacturing, and financial services





Use of AI in the World by Industries and Departments in 2023



Source:Statista



$$a\left(\frac{b}{c}\right) = \frac{ab}{c} \qquad f(x) \le 5$$

$$\frac{\left(\frac{a}{b}\right)}{c} = \frac{a}{bc} \qquad X^2 - 4X + 5 \le 5$$

$$\frac{n_1}{c} = \frac{a}{bc} \qquad X^2 - 4X \le 0$$

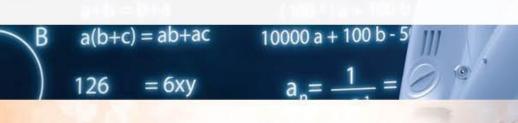
$$\frac{a}{b} = \frac{ac}{b} \qquad n(B \cap C) = 22$$

What is Machine learning (ML)?

- The process of using mathematical models of data to help a computer learn without direct instruction
- It is considered a subset of artificial intelligence (AI)

log

- ML learning uses algorithms to identify patterns within data, and those patterns are then used to create a data model that can make a prediction
- With increased data and experience, the results of ML are more accurate



= i a , a≥0 bi)+(c+di) = a-+bi)-(c+di) = a-(a+bi)(c+di) = ac-(a+bi)(a-bi) = a²+b² |a+bi| = a²+b²

 $+0_2 \rightleftharpoons 250_3$

Some key ML techniques



Supervised learning: Addressing datasets with labels or structure, data acts as a teacher and "trains" the machine, increasing in its ability to make a prediction or decision.



Unsupervised learning: Addressing datasets without any labels or structure, finding patterns and relationships by grouping data into clusters.



Reinforcement learning: A computer program helps determine outcome based upon a feedback loop.





Deep learning / neural networks:

Deep learning is an advanced type of ML that uses networks of algorithms that are inspired by the structure of the brain - neural networks. Deep learning typically requires a large data set to train on; training sets for deep learning are sometimes made up of millions of data points



Transformer model:

A transformer is a deep learning model that adopts the mechanism of self-attention, differentially weighing the significance of each part of the input data or tracking relationships in data, like word order in a sentence, but can be applied to other scenarios including fraud detection



Use of Big Data, AI, and ML in Financial Institutions

Application/Opportunities

Asset management

The use of AI and ML in asset management has the potential to increase the efficiency and accuracy of operational workflows, enhance performance, strengthen risk management, and improve the customer experience

Algorithmic Trading

- Al can be used in trading both to provide trading strategy suggestions and to power automated trading systems that make predictions, choose the course of action and execute trades
- AI-based trading systems operating in the market can identify and execute trades entirely on their own, without any human intervention
- Also assists traders in risk management

Sources: Shabsigh, G. (2023); Javaid (2024); Begenau et al.(2018); Lopez-Corleone et al.(2022)



Credit intermediation and assessment of creditworthiness

- Al-based models and big data are increasingly being used by banks and fintech lenders to assess the creditworthiness of prospective borrowers and make underwriting decisions, both functions at the core of finance
- In the context of credit scoring, ML models are used to predict borrowers' defaults with superior forecasting accuracy compared to standard statistical models

Integration of AI in Blockchain-based financial products

- Al algorithms can execute blockchain-based trades at high speeds and with precision, based on real-time data analysis and predictive models, improving trading efficiency and performance
- AI also analyzes blockchain data to offer insights and recommendations for optimizing investment portfolios, including those involving cryptocurrencies and blockchain assets

Sources: Shabsigh (2023); Javaid (2024); Begenau, et al. (2018); Lopez-Corleone et al. (2022)



Challenges and Potential Risks

Risk of tacit collusions

- Al and ML can cause similar trading behaviors among different entities due to the use of similar algorithms and data-driven strategies
- This can result in coordinated price movements and market impact without explicit agreements

Employment risks and the question of skills

- The deployment of AI and big data in finance requires different skills which only a few of the workers and leaders in finance have
- Firms will need to invest in human capital and acquire skills related to AI and Big Data management if they have to exploit value from data
- Al and ML technologies are likely to render some workers jobless as machines replace humans in automated processes

Sources: Shabsigh (2023); Javaid (2024); Begenau et al. (2018); Lopez-Corleone et al. (2022)



Use of Big Data, AI and ML in Central Banks



Monetary Policy and Economic Forecasting:

Big data and AI enable central banks to analyse vast amounts of financial and economic data for more accurate forecasting and decision-making in monetary policy.



Regulatory Compliance and Supervision:

Al and machine learning help in monitoring financial institutions for compliance with regulations, detecting potential risks, and improving oversight efficiency.

Source: Araujo et al. (2024); Doerr et al. (2021)



Globally

- The Bank of Canada: Since 2017, the bank has used AI to forecast inflation, economic activity, and demand for bank notes, track sentiments in key sectors of the economy, clean and verify regulatory data, improve efficiency and de-risk operations
- The European Central Bank (ECB): The ECB has used AI and Big data in communication, macroeconomic trends analysis, monitoring financial stability, assessing the effectiveness of monetary policy, and detecting systemic risks
- Federal Reserve (U.S.): Since 2011, the Federal Reserve has done away with the traditional supervision and forecasting methods. With modern technologies, the Federal Reserve uses ML models for economic forecasting, fraud detection, and monitoring of financial institutions. It also uses Al and Big data to analyze large datasets for better decision-making
- Bank of England: Since 2014, the Bank of England has been exploring the use of AI and ML in areas such as economic modeling, regulatory compliance, and financial stability monitoring

Source: Bank of England (2023), Bank of Canada- Macklem (2024), BIS Report (2024)

Africa

- The South African Reserve Bank: SARB uses Big Data and ML for fraud detection, financial stability monitoring, and enhancing its regulatory oversight. In 2022, the SARB published a working paper titled "Big Data Forecasting of South African Inflation," which shows the effort to incorporate these tools in forecasting, analysis and policy formulation(South African Reserve Bank)
- The Central Bank of Kenya (CBK) utilizes data from mobile money platforms to monitor economic activity and financial inclusion. Since the early 2020s, Kenya has also been exploring the potential of AI and ML in financial regulation and supervision to enhance financial services, reduce risk, and increase operational efficiency
- The Central Bank of Nigeria (CBN): Since the early 2020s has been incorporating AI, Big Data, and ML to enhance credit risk assessment, improve financial stability monitoring, and combat fraud and money laundering
- Concerning monetary policy, AI adoption is still in its early stages in many African central banks. There is growing interest in using AI to improve the accuracy and timeliness of policy analysis
- Sources:
- Bank of England: https://www.bankofengland.co.uk/prudential-regulation/publication/2022/october/artificial-intelligence
- Bank of Canada: Artificial intelligence, the economy and central banking Bank of Canada
- Bank of International Settlements: BIS Annual Economic Report 2024
- SARB: Big data forecasting of South African inflation
- Central Bank of Kenya:1598390763_Banking Sector Innovation Survey 2022.pdf

Selected Central Banks Using Big Data, Al and ML(Araujo, Doerr, Gambacorta and Tissot, 2024)

Name of Central Bank	Application Type
Bank Indonesia	Macro/financial analysis for monetary policy
Bank of France	Macro/financial analysis for monetary policy, and Supervision
Reserve Bank of Australia	Macro/financial analysis for monetary policy
Reserve Bank of New Zealand	Macro/financial analysis for monetary policy
Bank of Canada	Macro/financial analysis for monetary policy, and Payments oversight
Bank of Korea	Macro/financial analysis for monetary policy, and Payments oversight
Federal Reserve	Macro/financial analysis for monetary policy, and Supervision
Central Bank of India	Macro/financial analysis for monetary policy, and Payments oversight
Central Bank of Malaysia	Macro/financial analysis for monetary policy, and Supervision
Central Bank of Chile	Macro/financial analysis for monetary policy
Bank of England	Macro/financial analysis for monetary policy
Bank of Italy	Macro/financial analysis for monetary policy, Supervision, and Payments oversight
Czech National Bank	Macro/financial analysis for monetary policy
Bank of Thailand	Payments oversight
Central Bank of Ecuador	Payments oversight
Central Bank of Iceland	Payments oversight
Bank of Japan	Supervision
Bank of Spain	Supervision
Banco de Portugal	Supervision
Central Bank of Brazil	Supervision
Bank of the Republic of Austria	Supervision
Monetary Authority of Singapore	Supervision
South African Reserve Bank	Macro/financial analysis for monetary policy



Use of Big Data, Al and ML for Monetary Policy

Inflation Targeting:

- CBs can comprehensively analyze price movements and economic indicators, improving the ability to track and target inflation
- Al and ML tools can predict inflation trends by analyzing vast datasets and identifying patterns that may not be otherwise detected by traditional methods, helping central banks set more accurate inflation targets

Interest Rates Setting:

Al allows central banks to analyze economic data, including real-time financial transactions and market trends as well as model complex economic scenarios and predict the effects of interest rate changes on the economy

Communication of Monetary Policy:

Big data and AI ensure policy decisions are well understood by the public, and inflation expectations are kept well-anchored

Sources: Doerr, et al. (2021); Chiranjit and Joseph (2017); Cipollone (2024); Ozili (2024)

Exchange Rates Movements:

Big Data, AI, and ML help central banks monitor global economic conditions, as well as predict and analyze exchange rate fluctuations by processing large datasets from different sources; thus, helping informed decisions about currency interventions and policy adjustments

Money Supply:

Big Data, AI, and ML allow CBs to analyze spending patterns and money flows and predict demand and anomalies in financial behavior which can help central banks monitor and adjust money supply

Monetary Policy Transmission:

Al tools allow for real-time monitoring of economic indicators such as inflation rates, employment figures, and consumer spending as well as simulation of the monetary policy transmission mechanisms; hence, central banks can assess the immediate impact of monetary policy decisions and adjust the needed strategies

Sources: Doerr et al. (2021); Chiranjit and Joseph (2017); Cipollone (2024); Ozili (2024)

Forecasting Precision of ML methods (Yossi Shvimer, Victor Murinde and Avi Herbon, 2021)

Table 2: Out of sample results in 5 different models

Hybrid LSTM-ARIMA with sentiment 3.00681 1.7069

Hybrid LSTM-ARIMA without sentiment 3.01327 1.72781

ARIMA 3.18904 1.73033

LSTM with sentiment 10.8492 5.44829

LSTM without sentiment 10.997 5.71967

Table 2 shows that Hybrid LSTM-ARIMA with sentiment has the lowest RMSE score and MAE score

Apr 4 May 8 Jun 12 Jul 16 Aug 20 Sep 24 Oct

1d 5d 1m 5m 1y 5y

What are the Priorities?

1

Scenario Analysis:

Al models allow central banks to simulate various economic scenarios and assess the potential impacts of different policy decisions, helping policymakers to evaluate the effectiveness of their interventions under different conditions and choose the most appropriate course of action 2

Effective Communication Strategies:

Al can assist central banks in crafting clear and effective communication strategies by analysing how different messages are received by the public and financial markets, to enhance the credibility and transparency of monetary policy, leading to more stable market expectations

3

Stakeholder Engagement:

Big data enables central banks to engage more effectively with stakeholders, including the public, financial institutions, and government bodies; by understanding the concerns and expectations of these groups through data analysis, central banks can better align their policies with the needs of the economy.



A

Challenges and Limitations

Handling Big data is Resource Intensive:

Collecting and accessing information, infrastructure, high data security measure and high staff costs to clean data

Lack of Transparency:

Some AI and ML algorithms, especially deep learning models, are complex and operate like 'black boxes'. People need to know how these models arrive at specific decisions

Privacy and Security Issues:

As data collection grows, protecting personal financial information and ensuring its security become essential

Potential Biases:

Al systems may reinforce existing biases in data, potentially resulting in unjust decisions or prejudiced outcomes

Sources: Aldasoro et al. (2024) and OECD(2024)



Underlying Quality of Data may be low for public authorities like Central Banks:

Data might not be well structured, or measurements might introduce biases



Regulatory and Ethical Questions:

The fast-paced development of these technologies challenges current regulations and raises ethical concerns about their application



Reduce Operational resilience:

Over-reliance on AI to perform core tasks can weaken an organization's ability to withstand and recover from disruptions if AI systems fail or cause errors



Unethical Use:

Al models can be used to spread false information and data, facilitate fraud or launch cyberattacks



Central banks Risk Falling into "Echo-chamber' trap:

Al may become self-referential, repeating existing biases and increase the risk of forward guidance



Recommendations to Overcome Existing Challenges'...

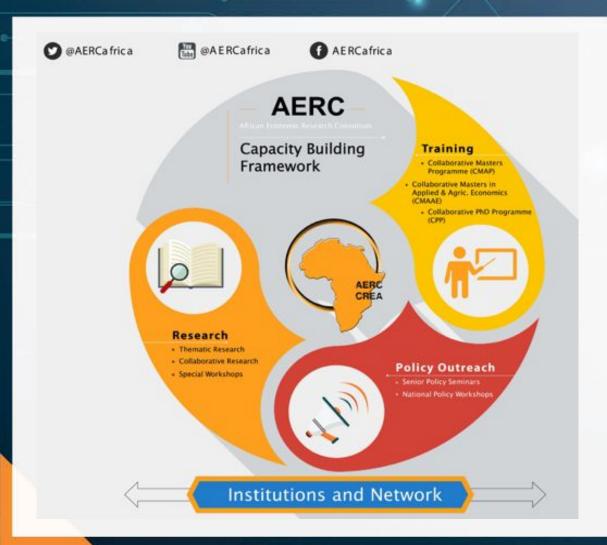
- 0 = 1
- Eliminate bias in AI models: Use diverse and representative data, conduct biased audits for biased outcomes, and involve diverse teams in the design and implementation of AI systems to enhance inclusivity
- Criteria and process of algorithms should be clear (improve interpretability of machine learning models) and be audited to enhance transparency to the public
- State-of-the-art security is required to ensure data safety and eliminate cyberattacks
- Reduce overdependence on AI as they are prone to errors and system failures
- Enhanced collaboration between regulatory bodies and AI developers as a crucial strategy for improving the resilience and transparency of financial systems
- Invest in AI Literacy and Training: Building expertise in AI and machine learning within central banks is essential to effectively implement and oversee AI technologies





Concluding Remarks

- The integration of Big Data, AI, and Machine Learning into monetary policy represents a transformative shift in how central banks and financial institutions operate; the technologies enable more accurate, real-time insights into responsive policy decisions
- By leveraging vast datasets and advanced algorithms, policymakers can better anticipate market trends, manage risks, and ensure financial stability
- However, the rapid pace of technological change also presents challenges, including the need for robust data governance, and the continuous upskilling of personnel
- The successful application of these technologies will be pivotal in shaping the future of monetary policy, driving innovation, and enhancing the effectiveness of monetary policy in an increasingly complex global landscape
- Nevertheless, humans need to stay in control of AI systems not only to guarantee their ethical use but also to handle accountability and sustain public confidence in these technologies; AI should complement (and not replace) human thinking



The AERC aims to prioritize Big Data, AI, and ML in its Strategic Plan 2025-2035, as an integral component of capacity building and knowledge generation in Africa

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