Beyond observational data – enhancing epidemiology analytics with crosscutting disciplines

HE²RO (Health Economics and Epidemiology Research Office)

Wits Health Consortium, University of Witwatersrand

27 September 2024







Meet the HE²RO team





Founded in 2003 in collaboration with Department of Global Health, Boston University



Using health economics, epidemiology, demography, modelling, big data and predictive data analytics, data management, behavioural sciences and quantitative and qualitative research methods.



Maintaining networks and relationships with DoH, Treasury and other government organisations



Developing strong local partnerships and leveraging international collaborations



Building and supporting local capacity development in a growing cadre of upcoming researchers

Sustained support for multi disciplinary research has allowed the South African national HIV treatment program to *respond* and adapt to the epidemic with evidence based policy change.



Health Economics and Epidemiology Research Office

Collaborations





Health Economics and Epidemiology Research Office







Symposium overview

Session	Presenter
Optimizing routinely collected data for impact	Oratile Mokgethi and Amy Huber
Costing of health interventions across COVID-19, TB and HIV	Nkgomeleng Lekodeba
Bridging modelling and health economics to inform the South African government's response to HIV and TB	Lise Jamieson
Machine learning and AI to support health: Your Choice and PREDICT	Caroline Govathson and Mhairi Maskew
Behavioural economics: an introduction to Indlela	Preethi Mistri
Panel discussion and interactive session	Presenters and panel members



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Optimizing routinely collected data for impact

Oratile Mokgethi and Amy Huber





What is AMBIT?

AMBIT: Evaluating differentiated service delivery for HIV in Africa

Objective: Describe **implementation scale**, estimate **benefits and costs**, and identify optimal **allocation of resources** for differentiated care in Africa

Data collection [SENTINEL]: Annual and ad hoc survey of sentinel healthcare facilities, providers, patients, and models

Data synthesis [GREAT]: Analysis of EMR and other routinely collected data

ADAPT: Optimization model



Who is AMBIT?

Health Economics and Epidemiology Research Office (HE²RO)





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Routine data



Many countries are scaling up electronic capture of medical information



Data from these EMR have the potential to answer important public health questions and inform policy and decision makers



However, routinely collected EMR data are not created for research purposes



Analyzing them requires knowledge of the limitations and sometimes creative solutions to the challenges



cords (n)

Client re



What obstacles are we facing?



Data completeness



Data quality



Data linkage

Data access, specifically to identifier



Data collected for routine monitoring and not epi studies



Segmented data systems



What are some methods to overcome these obstacles?

Data review and validation

Data linking

Target trial emulation



Data review and validation



DSD data field

- Distinguish b/w less vs. more intensive
- Is off-site included?



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EMR distribution among clinics

- Timing & distribution of EMR systems
- More complete EMR \rightarrow higher DSD uptake



Dispensing interval

• Is dispensing interval captured for everyone, including off-site DSD?

Laboratory records

- VL and CD4 tests broad vs. targeted
- EMR input via paper or linked database

Contextual issues with observational data

- DSD concurrent with UTT, COVID, DTG
- DSD enrollment not random



- Dynamics within cohort
 - If quality of care improves due to DSD,

clients may become eligible sooner





Data linking

- Clinical data linkage without identifiers is a useful tool for researchers and policymakers
- Without a unique identifier or routinely captured unique identifiers in the medical system, linking records can be difficult if name and DOB are not available.
- This linkage approach offers novel opportunities to evaluate South Africa's public sector HIV program





AMBIT SENTINEL survey and TIER.Net linkage





Target trial emulation method



Differentiated service delivery models (DSD) have been widely implemented.



Replacing conventional care with DSD models could benefit DSD clients and the back to end to en the health system.



DSD models have multiple potential benefits but prerequisite is sustaining good client outcomes.

~ —	
~ —	
~ —	

Compared outcomes of DSD clients to those eligible but not enrolled in DSD in South Africa.







Methods



Population and data source

• Electronic medical records (TIER.Net) for 18 facilities in 3 districts in South Africa



Target trial emulation methodology (TTE)

- TTE can assist in avoiding immortal time/survival bias in observational data
- First specify the protocol of the hypothetical randomized trial then use observational data to emulate each component of the target trial



Analysis

- Estimated risk differences (RDs) for retention in care and viral suppression
- Poisson model adjusted for age, gender, urban/rural status, province, WHO stage and CD4 count at ART initiation, as well as years on ART at trial enrolment
- Present RDs for both by target trial and pooled data



Target trial vs. emulation

Component	Target trial	Emulated trial
Eligibility	Age 18+; on ART >12 months; 2 suppressed viral	Same
	loads (<400 copies/ml); DSD naïve	
Intervention	DSD: enrolled in DSD model at baseline	Same
strategy	Non-DSD: remain in conventional care at baseline	
Intervention	Clients randomly assigned to DSD vs non-DSD	Clients non-randomly assigned to DSD model.
assignment		Randomization is emulated by adjusting for
		baseline covariates: age, sex, urban, province
Follow-up	Starts at randomization	DSD: follow-up starts at the first DSD visit in the
	Ends at earliest of death, LTF, 48 months.	trial baseline.
		Non-DSD: follow-up starts at the first visit in the
		trial baseline.
Outcomes	1) Retention in care and 2) Viral suppression	Same
	Each at 12, 24 and 36 months	
Causal	Intention-to-treat effect, i.e. effect of being	Observational equivalent of intention-to-treat
contrast	assigned to DSD vs non-DSD at baseline,	effect
	regardless of actual DSD enrolment.	
Statistical	Risk difference model estimating retention and	Same. Adjustment for baseline covariates did not
analysis	viral suppression in DSD vs non-DSD.	substantially change the estimates

Target trial for impact of DSD on retention/ suppression



Results

- Retention in care is similar or slightly higher for those in DSD models after 12, 24 and 36 months
- Proportion virally suppressed is similar for those in DSD models vs. those eligible for DSD (but not enrolled) at 12, 24 and 36 months

Non-DSD

DSD

 Estimates were robust across emulated target trials, and after adjusting for covariates



Limitations

- Target trial emulation assists with misaligned person time and eligibility criteria but residual confounding can remain
- Bias is very likely with regard to who was offered DSD enrolment
- Routinely collected EMR data limitations silent transfers, off-site medication collection not captured

Conclusions

- Model enrolment can be evaluated on the basis of other benefits (or costs), such as resource utilization or satisfaction - clinical outcomes are at least as good.
- Differentiated models for those not currently eligible for lower intensity care should be investigated
- Target trial emulation methodology is a robust method for defining a comparator arm and determining outcomes in routine care data

Key takeaways

- Limitations in the dataset often constrain the questions that can be asked
- There is value in analysing routinely collected EMR data for health outcomes, but caution must be applied
- Creating a working group with the Ministry/Department of Health, data base manager, and informatics team is a best practice
- Building in time for data review and feedback are essential
- Iterative feedback to the DoH/MoH strengthens analysis and ensures relevance of the work
- Observational data methods such as target trial emulation can assist

Acknowledgments

- Departments of Health in South Africa
- Implementing partners
- Sentinel site staff and clients
- Bill & Melinda Gates Foundation
- CQUIN, IAS, and other collaborating organizations
- AMBIT project team









Questions



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Costing of health interventions across COVID-19, TB and HIV: evidence based-research

Nkgomeleng Lekodeba







...a trip down memory lane: COVID-19 in South Africa

Fig. Number and cumulative number of laboratory-confirmed cases of COVID-19 by date of specimen collection, South Africa, 3 March 2020 – 25 March 2023



COVID-19: Cost of the Integrated Vaccination Programme in South Africa

- Ingredients-based, top-down costing at national, district and vaccine delivery channel level
- 4 hospitals, 47 primary healthcare facilities and 178 school-based delivery channels
- Provider (government) perspective; costs reported in 2022 USD

COVID-19 vaccine delivery via PHCs was cheaper per dose delivered and slightly higher for mobile outreach and more costly for school-based delivery channel

*National and district level cost allocated to each delivery channel; PHC- Primary Healthcare; costs includes vaccine procument costs





Source: Edoka et al 2024 BMC Health Services Research

COVID-19: Cost-effectiveness analysis of South Africa's vaccination programme

- Modelled costs, health outcomes of COVID-19 vaccination compared to a no vaccination scenario
- Estimated incremental cost-effectiveness ratio; cost per DALYs (disability adjusted life years) averted
- Provider perspective; costs reported in 2021 USD.

Table. Incremental costs, DALYs lost and Incremental cost-effectiveness ratio (probabilistic)- base case analysis

	Study period 1 (Feb-Nov 2021)		Study period 2 (Feb 2021-Jan 2023)			
	Total costs (US\$) DALYs lost		DALYs lost	Total costs (US\$)	DALYs lost	DALYs lost
		(Hospital death)	(Excess death)		(Hospital death)	(Excess death)
No vaccination programme	\$897,419,147	1,251,318	2,598,906	\$1,260,844,715	1,698,575	3,854,858
Vaccination programme	\$1,277,596,224	1,015,774	2,243,904	\$1,775,169,846	1,303,481	3,075,690
Difference	Incremental Costs	DALYs averted	DALYs averted	Incremental Costs	DALYs averted	DALYs averted
	\$380,177,077	235,544	355,002	\$514,325,131.00	395,094	779,168
ICER (US\$/DALY averted)		\$1,614	\$1,071		\$1,301	\$660

Source: Edoka et al..,2024 Vaccine



Vaccination programme became more cost-effective with inclusion of additional wave due to increased coverage, averting more deaths

COVID-19: Cost of adding rapid screening for diabetes, hypertension to COVID-19 vaccination queues in South Africa

- 3 public sector primary healthcare clinics and 1 academic hospital in South Africa, May-Dec 2022.
- Provider (government) perspective; bottom-up micro-costing; costs reported in 2022 USD.

Table. Median cost of NCD screening per patient outcome , 2022 USD (%)

Patient category (n)	Staff	Consumables	Diagnostic tests ²	Equipment	Overall median(USD, IQR)
DM screen positive (22)	\$5.50 (20)	\$2.89 (11)	\$18.98 (69)	\$0.01 (0)	\$29.62 (28.18, 31.50)
DM screen negative (1354)	\$1.54 (73)	\$0.57 (27)	\$0.00 (0)	\$0.01 (1)	\$1.69 (1.38, 2.49)
HTN screen positive (138)	\$3.41 (75)	\$0.65 (14)	\$0.48 (10)	\$0.02 (0)	\$3.53 (2.79, 4.62)
HTN screen negative (1248)	\$1.40 (61)	\$0.60 (26)	\$0.28 (12)	\$0.01 (1)	\$1.62 (1.36, 2.14)
DM + HTN screen positive (3)	\$7.94 (24)	\$3.25 (10)	\$21.98 (66)	\$0.02 (0)	\$31.57 (31.43, 34.12)
DM + HTN screen negative (1360)	\$1.35 (70)	\$0.57 (29)	\$0.00 (0)	\$0.01 (1)	\$1.62 (1.36, 2.11)

²Blood collection materials and lab handling fees

Source: Masuku et al..,2024 BMC Public Health

Incorporating NCD screening into ongoing mass health interventions such as vaccination programs could offer significant long-term benefits (e.g. early case detection)







Tuberculosis: Cost and cost-effectiveness of BPaL regimen used in drugresistant TB treatment in **South Africa** (and the Philippines)

- Treatment arms: BPaL 6 month current standard of care, 9–11-month standard short oral regimen (SSOR) and 18–21-month standard long oral regimen (SLOR)
- Patient and provider perspective (top-down & bottom-up micro-costing); reported in 2022 USD

Cost category (mean costs)	BPaL CAP (n=42)	SSOR (n=46)	SLOR (n=24)
TB diagnosis + monitor	\$61.5	\$66.7	\$79.9
Client-facing providers	\$382.8	\$296.7	\$341.6
Drugs (TB)	\$1,178.4	\$1,099.9	\$2,254.1
Drugs (ancillary)	\$3.8	\$31.7	\$32.5
Laboratory tests	\$44.2	\$35.5	\$54.7
Fixed costs/overhead	\$2,821.9	\$2,879.8	\$4,480.8
Equipment	\$150.7	\$153.7	\$239.7
Total cost per patient	\$4,643.2	\$4,563.8	\$7,483.2
Total cost per patient per outcome	N=39 (93%)	N=31 (67%)	N=10 (42%)
(favourable)	\$4,948.7	\$4,905.6	\$8,919.9
ICER (\$/favourable outcome)	\$311		Dominated

TABLE. Provider costs and cost-effectiveness of TB treatment provision stratified by outcome and treatment group (USD 2022 95% CI)

Evans.., et al 2024 (in press)

Bpal was cost-effective compared to SLOR and SSOR

Tuberculosis: Cost and cost-effectiveness of BPaL regimen used in drugresistant TB treatment in South Africa (and **the Philippines**)

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- Patient and provider perspective (top-down & bottom-up micro-costing); reported in 2022 USD

Cost category	BPaL (n=32)	SSOR (n=87)	SLOR (n=25)
TB monitoring	\$414.1	\$436.4	\$738.0
Client-facing providers	\$42.4	\$18.1	\$25.4
Drugs (TB)	\$568.3	\$408.3	\$1,686.3
Drugs (non-TB)	\$391.6	\$322.6	\$347.6
Laboratory tests/procedures	\$256.9	\$263.9	\$409.2
Fixed costs/overhead	\$320.0	\$317.3	\$403.8
Equipment	\$1.2	\$1.2	\$1.5
Total cost per mean (all)	\$1,994.5	\$1,794.9	\$3,611.7
Total cost per mean (favourable)	1,994.5	2,371.2	5,992.3
Cost-effectiveness			
ACER	1,995	\$3,122	\$10,032
ICER (favourable)		\$469	Dominated

TABLE. Provider costs and cost-effectiveness of TB treatment provision (USD 2022 95% CI)

Source: Evans *et al..,* 2024 IJTLD OPEN

Total provider costs per successful treatment were lowest with BPaL but more costly for SSOR and SLOR

National TB Cost model

Generating inputs needed for the TB Investment case

- Systematic review of primary cost data for TB interventions in South Africa
- Existing interventions: augmented bottom-up cost analyses
- Novel interventions: ingredient-based cost analyses
- Provider perspective (e.g. South African government)
- Target populations estimated using the Thembisa TB model
- Total cost of the TB programme = average cost x target population



Intervention	USD 2024
IPT for PLHIV/HH contacts	\$10.55
ЗНР	\$22.19
Symptom screening (PLHIV)	\$0.69
Symptom screening (General PHC)	\$0.69
Symptom screening (HH contacts)	\$1.38
Symptom screening (Door-to-door)	\$1.57
Mobile chest radiography	\$40.36
Xpert MTB/RIF Ultra	\$16.36
Culture in liquid medium (Xpert neg PLHIV)	\$5.36
Smear microscopy	\$4.50
TUTT for PLHIV/previous TB	\$18.98
Outpatient treatment (DS-TB), adults	\$123.36
Inpatient treatment (DS-TB), adults	\$295.48
Treatment (DR-TB), adults	\$860.32
Outpatient treatment (DS-TB), children	\$98.61

In-patient is more expensive as those diagnosed in hospitals typically present with advanced TB disease requiring more complex treatment







HIV: Comparative cost and cost-effectiveness analysis of viral load testing modality in South Africa

- Bottom-up (PHC), top-down (testing hub) costing approaches using decision analytical model
- Cost per successful test completed compared to conventional testing
- Provider perspective and reported in 2023 USD



Fig. Costs per successful test completed by modality

Table. Cost-effectiveness of modalities that increase the number of tests successfully completed

	Tests	Total	
Scenario	completed	costs	ICER
eLABS WITH order entry & HPRS	449,697	\$11,541,424	*
eLABS withOUT order entry	437,095	\$11,885,001	Strongly Dominated
Conventional care	433,647	\$11,906,729	Strongly Dominated
eLABS WITH order entry	448,478	\$12,095,223	Strongly Dominated

Source: Lekodeba et al.., (in draft)

eLABS WITH order entry & HPRS was cost-effective

★ Cost-effective
HIV: Optimal mix of differentiated service delivery models for HIV treatment in Zambia: a mathematical modelling study

- Excel-based model; retention and viral suppression rates from national cohort on ART (≥15 years), 2018-2022
- Evaluated the incremental cost per additional ART client virally suppressed on treatment vs the base case.
- Costs were estimated from the previously-published data, provider perspective, reported in 2023 USD

Scenarios	Total number of people retained (n, % change	Total number of people suppressed (n, % change	Total health system cost (USD, % change)	ICER per additional person suppressed	
	compared to base case)	compared to base case)		(USD)†	
Base case	817,948 (n/a)	770,086 (n/a)	\$84,332,234 (n/a)	n/a	
6MMD	827,415 (1.2%)	782,545 (1.6%)	\$83,095,136 (-1.5%)	Cost-saving	
				(vs base case)	
6MMD & AGs	831,498 (1.7%)	791,712 (2.8%)	\$85,336,549 (1.2%)	\$245	
AGs	835,581 (2.2%)	800,878 (4.0%)	\$87,577,961 (3.8%)	\$245	
FTRs & AGs	838,225 (2.5%)	802,755 (4.2%)	\$88,483,273 (4.9%)	\$482	
FTRs	840,869 (2.8%)	804,632 (4.5%)	\$89,388,584 (6.0%)	\$482	
AGs & HAD	844,609 (3.3%)	807,356 (4.8%)	\$109,574,215(29.9%)	\$7,409	

Table. Health outcomes, health system costs, and ICERs for the DSD scenarios on the cost-effectiveness frontier

Source: Lekodeba et al.., 2024 (in-press)

Results: Optimal mix of differentiated service delivery models for HIV treatment in Zambia: a mathematical modelling study

Fig. Total number of people suppressed on treatment by total health system costs for scenarios on the cost-effectiveness frontier



Source: Lekodeba et al..., 2024 (in-press)

Conclusions

- Economic evaluation provides useful tools and results for evidencebased decision-making that may help enhance outcomes
- This process relies on identifying:
 - all relevant alternatives
 - appropriate methods for measuring and valuing costs and benefits
- Accurate costing helps inform government planning and budgets
- Implementing interventions that are not appropriately evaluated may lead to incorrect recommendations on what to implement and for whom



Questions



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Turning data into impact: Bridging modelling and health economics to inform the South African government's response to HIV and TB

Lise Jamieson





Why South Africa



Country with the largest number of people with HIV and the largest HIV programme

- 20% of the global population living with HIV
- 20% of the global population on ART
- 20% of the global population who have initiated PrEP

TB incidence is amongst the highest in the world at >500 per 100,000 population (WHO, 2021)

• TB/HIV co-infection prevalence the highest in the African region (UNAIDS, 2021)

South Africa's HIV/TB domestically funded:

- South African government (69% of total expenditure)
- PEPFAR (24%)
- Global Fund (2%)







The Thembisa HIV model



- More information available on: www.thembisa.org
- Deterministic compartmental mathematical model of the South African HIV epidemic, designed to answer policy questions relating to HIV
- Also a demographic projection model and a source of demographic statistics
- Used by UNAIDS and the SA Department of Health in global reporting of HIV indicators
- In Thembisa, the population is stratified by demographic characteristics (age, sex), sexual risk behaviour, HIV prevention access and HIV disease stage and treatment
- Recently the model has been extended to include tuberculosis



Data sources used in calibration

Data source	National	Provincial
Antenatal surveys: HIV prevalence	\checkmark	\checkmark
Household surveys: HIV prevalence data	\checkmark	\checkmark
Household surveys: ever tested for HIV	\checkmark	
Household surveys: ARV metabolites	\checkmark	\checkmark
Key population surveys (FSWs, MSM): HIV prevalence	\checkmark	\checkmark
Recorded deaths: adults	\checkmark	\checkmark
Recorded deaths: children	\checkmark	
DHIS: number of adults and children on ART		\checkmark
DHIS: HIV prevalence in adults, children tested for HIV	\checkmark	
NHLS: age distribution of children on ART	\checkmark	\checkmark
TIER: age distribution of adults on ART	\checkmark	\checkmark
Child PIP: HIV diagnosis in child deaths	\checkmark	





Optimised interventions and targets

The HIV Investment Case

- Iterative exercise to establish the most cost effective mix of interventions and optimal coverage under country's HIV budget (allocative efficiency)
- Optimisation routine
 - considers cost-effectiveness of each intervention (cost per life year saved)
 - iteratively adds the most cost effective intervention to a rolling baseline
- Constrained: mix of interventions given current budget
- Unconstrained: scales up interventions regardless of funding envelope
- Cost and impact over a 20 year period
- 2023 HIV Investment Case report:
 https://www.berezz.org/publications/south.afr

https://www.heroza.org/publications/south-african-hiv-investment-case/



HIV Investment Case 2023: optimisation results

Intervention (scaled-up coverage)	Cost per life year saved (ZAR)				
CONSTRAINED SCENARIO (affordable under current budget)					
Condom distribution (1.3bn/year)	Cost-saving				
HIVST optimized package (500,000 tests /year)	26,123				
Infant PCR testing at 10 weeks (95%)	27,900				
ART (81% monthly linkage)	42,263				
UNCONSTRAINED SCENARIO (interventions included in add	dition to above)				
ART (95% monthly linkage)	45,427				
HIVST optimized package (1m tests /year)	69,426				
HIV rapid antibody testing at 18 months (95%)	73,524				
Infant PCR testing at 6 months (95%)	107,122				
HTS general population (17m tests/year)	107,767				
HTS adolescents (95%)	190,583				
MMC (600,000 circumcisions/year)	482,978				
Oral PrEP package (30% FSW/MSM; 15% AGYW)	1,230,638				
CAB-LA package, max duration (64% FSW/MSM; 30%					
AGYW)	1,312,501				
CAB-LA for pregnant women (18%)	2,150,040				
Oral PrEP for pregnant women (14%)	2,239,265				
EIMC (70%)	566,585,885				

- Cost effectiveness measure: incremental cost (ZAR) per life year saved (LYS) of increasing coverage beyond baseline coverage
- Increasing condom provision is costsaving
- Next most cost-effective would be to scale up:
 - HIVST to 500k tests/year
 - Infant PCR testing at 10wks to 95%
 - ART monthly linkage to 81%
- All other interventions thereafter are not affordable to scale up under current budget, but remain affordable at baseline coverage
- Report:

https://www.heroza.org/publications/s outh-african-hiv-investment-case

The Thembisa TB model

- Separate annual rates of health seeking for people with TB symptoms, and people in the general population
 - Health seeking also differs by sex, HIV status
- People with symptoms at PHC get microbiologically tested at rates that depend on their HIV status, sex and reason for seeking care. Rates change over time
- Model allows for changes over time in the sensitivity and specificity of the lab testing algorithm (taking into account changing guidelines, imperfect uptake)
- Not all diagnosed individuals start treatment (initial LTFU). And people without TB can start treatment (false positives)
- Model allows for empirical treatment.





Current trajectory of TB epidemic and programme interventions (no changes) Informed by stakeholders: NSP, NTP-SP/ TB Think Tank, etc Aggressive efforts to reduce TB incidence and mortality to meet WHO's End TB targets



TB mortality impact of scenarios





Total cost by scenario 2023-2027 [2022 ZAR]



Modelling to Inform HIV Programmes in Sub-Saharan Africa (MIHPSA)



- The HIV Modelling Consortium is a network of epidemiologists, mathematical modellers, health economists, policy-makers and other stakeholders
- MIHPSA Collaboration supports the development and implementation of mathematical modelling to inform policy decision-making for HIV programmes in sub-Saharan Africa: South Africa, Zimbabwe and Malawi
- Aims to pool insights from multiple models with close engagement from stakeholders in order to inform the future design of country HIV programmes in African countries
- Learn more: http://hivmodeling.org/mihpsa



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Questions



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Machine learning and AI to support health

Caroline Govathson and Mhairi Maskew









Your Choice: Reducing Stigma in HIV Risk Assessment

27 September 2024





HIV in South Africa: The ongoing challenge and the need for tailored prevention efforts

Where we are:

8,5 million people
living with HIV189,000 people
(64% women)
were newly infected
with in 202245% of women newly
infected were <24yrs
High risk populations
being left behind

To provide appropriate, effective HIV care we need meaningful conversations, but there remain significant barriers for both clients and healthcare providers



To provide appropriate, effective HIV care we need meaningful conversations, but there remain significant barriers for both clients and healthcare providers





"You ask because some of the people, you know they are sex workers, but they won't tell you, you understand, when they come to the clinic, a person won't come to the clinic and tell you that I'm a sex worker"





A tech company and behavioural science unit used AI to try and solve this by leveraging LLMs to act as a conversational agent bridging the gap between client and healthcare provider



Use case tested: Utilise a LLM powered conversational phone app to gather data, estimate HIV vulnerability for clients, and provide a tailored summary that could be shared with a provider and support development of an HIV prevention plan.



In less than a year, we collaboratively designed the prototype for an AI powered HIV counsellor, obtained feedback from clients and healthcare providers based on hands on experience of the app and its data



Preliminary results show high (>90%) usability, acceptability and appropriateness



Continual, iterative design of app using multiple platforms to ensure safety for testing Enrolled 130 clients and assessed the usability, acceptability and appropriateness of the app for HIV risk screening Enrolled 20 health care providers



The key take away is that AI has a powerful role to play in the future delivery of healthcare, but there are areas of concern and significant need for further development

1	Conversation Accuracy : Client info & LLM answers	 Recipients of Care are willing to have an open and frank conversation ClaudeAI was marginally better than ChatGPT for conversation accuracy LLMs are not always delivering accurate information
2	Assessing HIV Vulnerability : Clinician summaries & LLM style	 LLMs can summarize relevant info for clinicians ClaudeAI scored higher than ChatGPT for the summary output LLMs struggle with local context, culture, slang, and tone
3	Technical Support: LLM needs ongoing iteration and support from dev team	 Application proved stable on low-cost smartphones LLMs required ongoing tweaking and support based on real-time feedback to to the development team
4	Intention-Action Gap: Users engaged with LLM about access to PrEP	 LLM uses language that promotes trust, builds rapport and empowers agency for behaviour change LLMs don't always have the appropriate language or nuanced understanding

The transcripts of app conversations showed active engagement; clients often had specific questions of their own related to HIV acquisition which the LLM answered but at the same time the LLM also steered the conversation to gather data to determine likelihood of HIV vulnerability.

"Yes I would like to know my status, although I'm anxious because I've once slept with someone who is HIV positive" "If my partner have sex with someone and come to me without taking a shower what will happen to me?"

Average conversation length was 43 user interactions and 48 with audio-to-text function

Overall internal accuracy rating was above 3.5 out of 5!

(<u>Not validated</u>: 1=very poor, inaccurate, dangerous, misleading information; 5=very good, fully accurate, relevant, very low risk of misinformation) ClaudeAl rated higher for conversation and summary





Insights from client and healthcare provider interviews on app's ability to address challenges



Health care providers

Summary of long conversations **Client autonomy** Safe space, honest conversation Judgment free zone Time to engage and ask Space to discuss other issues Flags other health issues

IMPLEMENTATION BARRIERS

<u>ה</u> 1

Resistance to change | Infrastructure | Language | Other conditions

Clients humanized the LLM, confided in the app and asked intimate questions. The LLM was able to accurately address basic clinical questions, showed empathy and offered practical suggestions

User: Thanks again its was nice talking with you

- LLM: You're very welcome. I'm glad we were able to have this compassionate discussion about such an important topic.[...] User: *How I going to get u I will love to chat with you* again
- LLM: I'm afraid I'm not a real person that you can contact directly....
- LLM: [...] Have you been tested for HIV in the past? Knowing your status can affect the discussion. Please let me know if a test result is something you'd like guidance on as well. User: *No, because eish am little bit scared* LLM: I understand feeling scared, getting tested can bring up worries. [...] If you'd like, we can look at resources to help overcome fears.



User: How does one use PrEP and how sure are you that it's 100 percent

LLM: PrEP, or pre-exposure prophylaxis, is a way for people who do not have HIV ...to prevent HIV infection by taking a pill every day. [...] When taken daily as prescribed..... 99% when used consistently. ...PrEP works best when combined with condoms.

User: So tell me, there is this new style of sex call rimming so I wanted to check is that thing safe to have LLM: Rimming, or oral-anal contact, can carry a risk of transmitting certain infections [...] If you have any specific concerns or questions about rimming or other sexual activities, it's best to consult with a healthcare provider who can provide personalized guidance. Healthcare providers were aware of the issues that their clients were struggling with and saw promise in the app for addressing some of these; they did raise some concerns about the use of technology in their clinics.

"Sometimes, yes. They're afraid of the healthcare providers, you know saying or voicing out their challenges and yes. So, we find that some of them don't divulge some of the information unless you dig and talk, yes."

"You ask because some of the people, you know they are sex workers, but they won't tell you, you understand, when they come to the clinic, a person won't come to the clinic and tell you that I'm a sex worker"

"..here are no printers so definitely I won't even talk about printout. Let's say we have PC
– cheaper stuff. I won't talk about tablets, something that will be more difficult to...security-wise"



Where to next? "IMPACTFUL SCALING"

This app can help us identify and connect individuals at substantial risk of HIV infection to tailored prevention strategies, including PrEP

90% of people at risk are linked to people centred, integrated services

95% of those at risk use prevention (including PrEP) Less than 10%

experience stigma and discrimination

The UNAIDS 2025 Targets for the HIV response.

This is just the beginning!

With our scalable learnings & tech platform we can leverage this across a breadth of healthcare conversational use cases with similar limitations to HIV prevention

(e.g. TB, STIs, diabetes and hypertension)



Applying machine learning to routine HIV patient data to predict missed clinic visits:

The PREDICT study



What methods and tools are needed to predict risk of treatment interruptions?

- PREDICT model
- Routinely collected EMR data from:
 - >310,000 ART clients
 - Districts in Mpumalanga and Free State



- Model predicted risk score for IIT for each observed visit using machine learning approaches
- Tested against known visit outcomes
- Predicted 2 of 3 treatment interruptions



How can risk scores be used in clinic settings?

- Threshold approach similar to casualty/field triaging
- Groups are segmented based on a visit-based risk score rather than condition urgency or perceived priority of need for intervention





Can we understand drivers of risk?

	n	IIT%	vs ADULT FEMALES			
Group by demographics only			Absolute risk difference	RR	95% CI	
ADULT FEMALES	4,855	13.4%	-	1.00	-	
AGYW	496	16.0%	3%	1.17	0.94 - 1.45	
ADULT MALES	176	13.0%	0%	0.97	0.66 - 1.44	
АВҮМ	2,960	14.0%	1%	1.01	0.9 - 1.13	



Can we understand drivers of risk?

			vs ADULT FEMALES			
Group by behaviour only	n	IIT%	Absolute risk difference	OR	RR	95% CI
ADULT FEMALES	4 920	13.3%	-	1	1.00	-
RETURNING INTERRUPTER	1,123	20.0%	0.07	1.68	1.55	1.35 - 1.78
LATE TWICE	999	18.0%	0.05	1.51	1.41	1.21 - 1.65
PREPARED AND LATE	551	18.0%	0.05	1.47	1.38	1.14 - 1.68
LIVE CLOSE ALWAYS LATE	1,138	17.0%	0.04	1.39	1.33	1.13 - 1.55
PROMPT & LOYAL	1,593	15.0%	0.02	1.22	1.19	1.03 - 1.36
DISILLUSIONED DISCLOSERS	1,394	15.0%	0.02	1.2	1.17	1.02 - 1.34
LONE RANGER	1,739	14.0%	0.01	1.12	1.11	0.96 - 1.27
UNEXPECTED UNSUPPORTED	964	14.0%	0.01	1.11	1.1	0.92 - 1.31
PRIOR TEST AND PROMPT	2,013	13.0%	0	0.97	0.97	0.85 - 1.12
EMPLOYED YOUTH PAYDAY	347	13.0%	0	0.95	0.96	0.73 - 1.26
SUPER GREEN	2,739	11.0%	-0.02	0.69	0.72	0.63 - 0.83


Can we understand drivers of risk?

Group by demographics AND				vs ADULT FEMALES			
behaviour		Don%	UT9/	Abc		DD	05% CI
	1 0 2 0	FOP%	12 20/	ADS	UR	1.00	95% CI
ADULI FEIVIALES	4 920	51%	13.3%	-	1	1.00	-
ABYM RETURNING INTERRUPTER	23	0.3%	34.8%	0.214	3.44	2.59	1.473 - 4.555
ABYM LATE TWICE	21	0.2%	23.8%	0.104	2.01	1.77	0.822 - 3.823
AGYW LATE TWICE	61	0.7%	23.0%	0.096	1.92	1.71	1.074 - 2.719
AGYW RETURNING INTERRUPTER	50	0.6%	22.0%	0.086	1.82	1.64	0.967 - 2.776
ADULT FEMALES RETURNING							
INTERRUPTER	651	7.7%	18.9%	0.055	1.62	1.50	1.256 - 1.795
ADULT MALES RETURNING							
INTERRUPTER	399	4.7%	19.3%	0.059	1.54	1.44	1.161 - 1.779
ADULT FEMALES LATE TWICE	563	6.6%	17.6%	0.042	1.44	1.37	1.124 - 1.657
ADULT MALES LATE TWICE	354	4.2%	18.1%	0.047	1.42	1.35	1.066 - 1.700
AGYW PROMPT LOYAL	119	1.4%	16.8%	0.034	1.3	1.25	0.834 - 1.878
ADULT MALES PROMPT LOYAL	548	6.5%	15.9%	0.025	1.22	1.18	0.962 - 1.452
ADULT FEMALES PROMPT LOYAL	895	10.5%	15.2%	0.018	1.2	1.17	0.979 - 1.388
ABYM PROMPT LOYAL	31	0.4%	9.7%	-0.037	0.69	0.72	0.245 - 2.118

Can we understand drivers of risk?

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behaviour	n	Рор%	IIT%	Abs	OR	RR	95% CI
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How can this help us intervene?

- We can pair combined demographic and behavioral profiles with interventions that are relevant, timely, and responsive to individual needs
- Start client-HCW conversations centered on client choice and needs





Questions



WCE 2024 HE2RO symposium

WhatsApp group









Introduction to Indlela

Preethi Mistri





Background



Image courtesy of "Living on a prayer! Navigating a safe passage to health care in the EC". Spotlight, 11th November 2019 | Alicestine October

- "Last mile" challenges in achieving the 95-95-95 targets may lie in human behaviour
- Behavioural economics provides a unique lens to identify and design low-cost behavioural solutions ("nudges")





Mission statement

Building capacity to design and test nudges and other behavioural solutions to improve the effectiveness of health services and achieve better health outcomes in South Africa and regionally





Hope Shateyi Data Manager

Busisiwe Sibiya Indlela Fellow

Onthatile Maboa Indlela Fellow

Co-Director



Behavioural Science Scientific Director



Brendan Maughan- Brown Behavioural Science Technical Behavioural Science Technical Expert



Caroline Govathson BE Fellowship Lead and Country Lead for Zimbabwe

Lawrence Long

Expert



Laura Rossouw

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(Data Analyst)

Neo Ndlovu B-Hub Lead and Country Co-lead for Kenya

Nomsa Mahlalela Nudge Associate



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Dino Rech





Francois Venter Ezintsha

Ingrid Katz PEPFAR



U.S. Centers for Disease Control

and Prevention (CDC)

International AIDS Society (IAS)



United States Agency for

International Development

Audere





Joshua Graff Zivin International Economic Relations at UC San Diego

Justine Burns University of Cape Town (UCT) RUBEN









Mohan Balachandran Center for Healthcare Transformation and Innovation Mosima Mabunda Discovery/Vitality



Romy Overmeyer Health (DOH)



The Aurum Institute





Zwebuka Khumalo South African National AIDS Council (SANAC)



Linda-Gail Bekker

South Africa Department of



Solange Baptiste International Treatment





















EAB

Preethi Mistri

Simamkele Bokolo Communications Lead and Local Technical Design Lead and Country Lead for Malawi





What we do?

Nudge Framework





A practical guide to behavioural design based on the NUDGE framework









Indlela's B-Hub and B-Lab



Behavioural Hub (B-Hub)

Activities:

- Recruitment
- Experimentation
- Digital tool testing



Activities:

- Behavioural Insight Tests
- Contextual inquiry
- Design workshops





Indlela's B-Hub and B-Lab



WORLD CONGRESS OF EPIDEMIOLOGY 20





Behavioural Insights Tests (BITs)

UNIVERSITY OF THE WITWATERSRAND.



WORLD CONGRESS OF EPIDEMIOLOGY



Indlela Behavioural Insights Test - PREDICT

	1. For each question, circle an answer and add the points in the "score" column. Sum all scores into "Total".					
		Adherence Scorecard	0	1	3	Score
		What is the client's age group?	Young adult (18-35)	Adult (36-59)	Senior (60+) Child (0 -17)	1
		For today's visit, is the client:	Late	First visit	On Time Early	3
it File		For their last visit, was the client:	Unknown	First or second visit	On Time Early	0
Clien		Has the client ever been over a month late?	More than once	Once	Never	1
		When was the client's last visit?	5 or more months ago	3-4 months ago First visit	0-2 months ago CCMDD or Fast Lane	3
		How many times has the client ever visited this facility?	0 - 4 visits	5 - 10 visits	11 or more visits	0
	7	Have you disclosed your HIV status to your friends or family?	No disclosure Partial	Full disclosure		1
osocia		How much time did it take you to get here?	More than 30 mins	30 mins or less		1
Psych	9	How many people do you live with?	Other number	2 - 6 other people		1
		Are you employed or studying?	No	Yes		1
			Total Adheses	0	Total	12
	0	0 1 2 3 4 5 6 7 8 9 10	11 12 1	3 14 15 16	17 18 19 20	21 22
	low-score		mid	-score	high-score	



Treatment referral plan

FOR BETTER HEALTH

Please note that suggested strategies must be recommended in line with patient eligibility according to the national guidelines.

Folder number

Adherence Score 12345678910 111213141516 171819202122

Based on our conversation, I am referring you to the following support to help you attend all your visits (You can choose more than one intervention):

Today's date

	Decant patients		Adherence counselling
	3 month repeat script Space and fast lane Choose appointment		Disclosure assistance Refer to CSTO/social worker for adherence Refer to facility case manager
	Other		
ealth c	are provider initials:	Patient initials	



Indlela B-Hub test – Your Choice "YOUR CHOICE" **Behavioural design & research Solution design Jan-April** July '23 Aug-Sept '23 Oct-Dec '23 **'24** Ś **#** audere **BOSTON** UNIVERSITY Penn UNIVERSITY OF THE WITWATERSRAND.



Regional expansion: Kenya, Zambia, Zimbabwe











KAVI Institute of Clinical Research UNIVERSITY OF NAIROBI

















Uniting against Poverty







REPUBLIC OF ZAMBIA MINISTRY OF HEALTH



Dissemination/ publication/ communications, platforms

hipo"; Mutasa, Barry"; Rees, Kate*.8

Webinars, symposia & publications





Website & Community of Practice



Our quarterly newsletter shares the latest news from Indlela as we build a community of practice around the use of behavioural economics and nudges to improve uptake and delivery health services in South Africa and the region. We hope this newsletter is helpful for you - please do share any feedback or suggestions that you may have. To sign up to receive our quarterly newsletter click here.





Thank You

contact us at Indlela@heroza.org









www







Questions



WCE 2024 HE2RO symposium

WhatsApp group



