

Hyperlocal Targeting of Vaccine Hesitancy in Malawi

Prepared by Fraym for Johnson & Johnson Global Public Health

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Hyperlocal Support of Vaccine Uptake: Overview

We aim to bring local understanding of uptake barriers across SSA

Goals



A **deeper understanding** of the 3C's occur locally and across entire countries to inform broad Risk Communication and Community Engagement (RCCE) efforts.



A detailed mapping of J&J's consumer segments and media consumption patterns across the country to close the gap between data and action.



An interactive tool to equip implementors with hyperlocal data to overcome barriers to vaccine uptake faster.

Outputs



Comprehensive reports containing overviews and detailed assessments of hyperlocal patterns of vaccine confidence, convenience, complacency, consumer segmentation, and media consumption patterns across the entire country.



DATAfraym® – an interactive web-based dashboard – access, for custom data exploration, analysis, and exports, with mapping available at a 1 km² level of granularity.

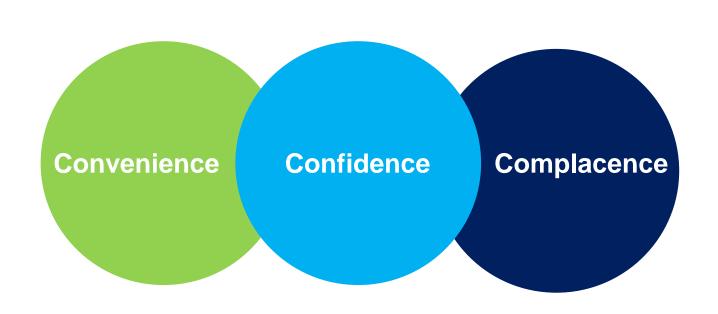


Complete datasets available regarding vaccine confidence, complacency, convenience, consumer segmentation, and media consumption patterns across the entire country, at a 1 km² level of granularity.





Paired with World Health Organization's 3Cs Framework, Fraym has mapped J&J's Consumer Segments to identify where vaccine uptake challenges are likely to occur

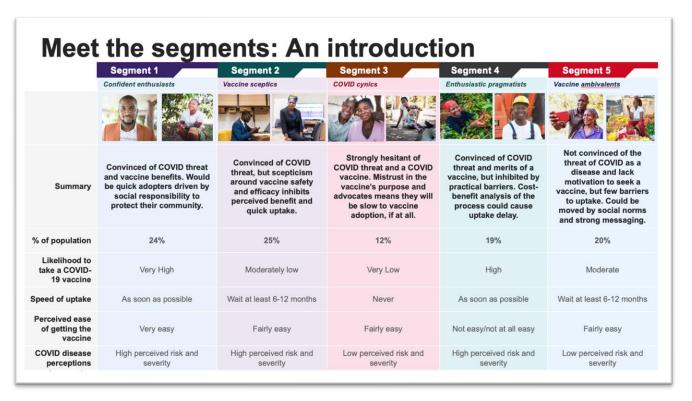


The 3Cs

Confidence: High confidence = *More* likely to take a vaccine

Convenience: High convenience = *More* likely to take a vaccine

Complacency: High complacency = *Less* likely to take a vaccine



Consumer Segmentation

Different segments of people have different motivations and reasons to not get a COVID- 19 vaccine (barriers)

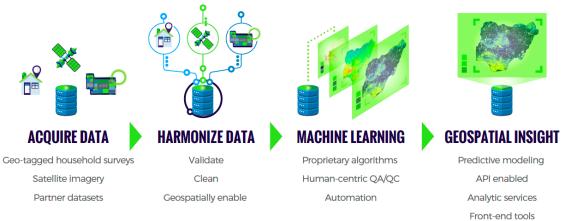


Methodology

Machine Learning for Hyperlocal Mapping

Fraym has built machine learning (ML) software that weaves together geo-tagged household survey data with satellite imagery to create localized population information (1 km²).

- The primary ML model input is data from high-quality, geo-tagged household surveys. Key indications of a high-quality household survey include implementing organization(s), sample design, sample size, and response rates. After data collection, *post-hoc* sampling weights are created to account for any oversampling and ensure representativeness.
- The second major data input is satellite imagery and related derived data products, including Earth observation (EO) data, gridded population information (e.g., human settlement mapping, etc.), proximity to physical locations (e.g., health clinics, ports, roads, etc.) and biophysical surfaces like soil characteristics. As with the survey data, Fraym data scientists ensure that the software only uses high-quality imagery and derivative inputs.
- To create spatial layers from household survey data, Fraym leverages machine learning to predict an indicator of interest at a 1 square kilometer resolution. This methodology builds upon existing, tested methodologies for interpolation of spatial data. The resulting model is used to predict the survey data for all non-enumerated areas. A similar approach was originally developed by academic researchers focused on health outcomes, which were expanded upon by USAID's Demographic and Health Surveys program since then by Fraym and others.¹



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Note 1: Gething, Peter, Andy Tatem, Tom Bird, and Clara R. Burgert-Brucker. 2015. Creating Spatial Interpolation Surfaces with DHS Data DHS Spatial Analysis Reports No. 11. Rockville, Maryland, USA: ICF International. Other notable, relevant work includes: Weiss DJ, Lucas TCD, Nguyen M, etal.Mapping the global prevalence, incidence, and mortality of *Plasmodium falciparum*, 2000–17: a spatial and temporal modelling study.Lancet 2019 and Tatem A, Gething P, Pezzulo C, Weiss D, and Bhatt S. 2014. Final Report: Development of High-Resolution Gridded Poverty Surfaces. University of Southampton. https://www.worldpop.org/resources/docs/pdf/Poverty-mapping-report.pdf

3Cs Model of Vaccine Hesitancy

Fraym used the WHO Sage Working Group on Vaccine Hesitancy backed 3Cs Model to analyze and segment vaccine hesitant populations.

The 3Cs Model of Vaccine Hesitancy

- Confidence: Trust in the effectiveness and safety of vaccines, the system that delivers them, and the motivations of policymakers.
- Convenience: Physical availability, affordability, and willingness-to-pay, geographical accessibility, ability to understand, and appeal of immunization services.
- Complacency: Level to which perceived risks of vaccine-preventable diseases are low and vaccination is not deemed a necessary preventative action.

Source 1: https://www.who.int/immunization/sage/meetings/2014/october/1_Report_WORKING_GROUP_vaccine_hesitancy_final.pdf

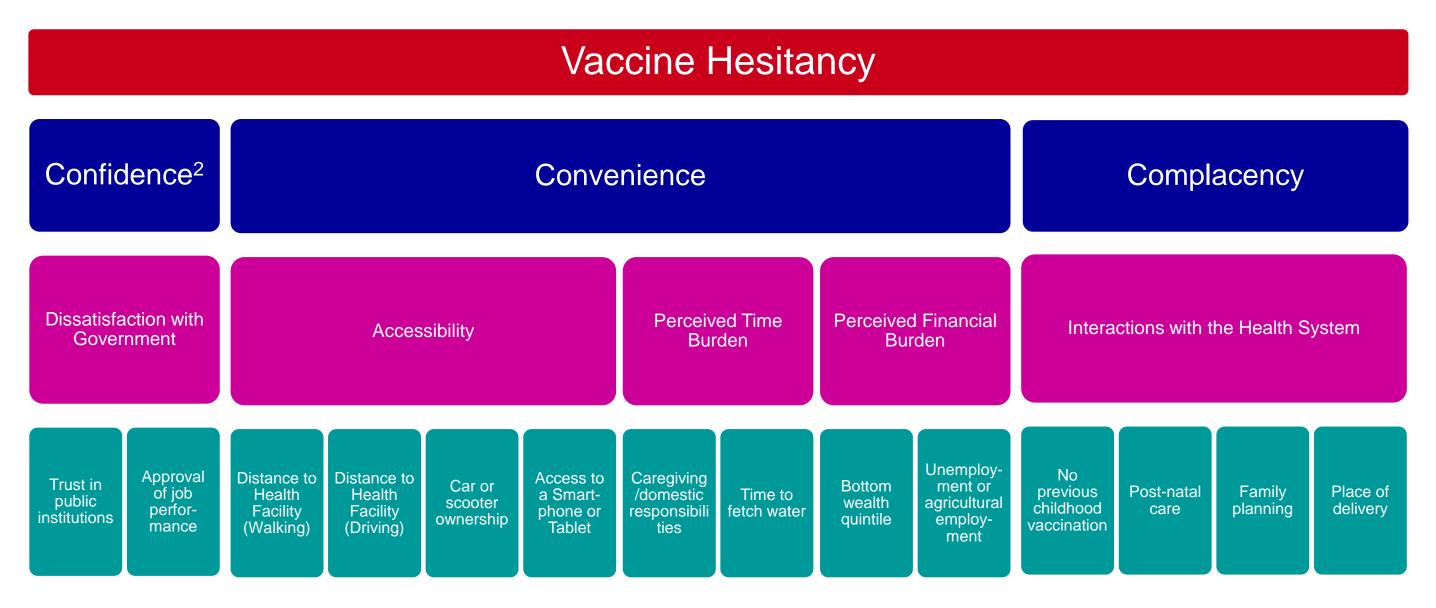
Source 2: https://africacdc.org/download/covid-19-vaccine-perceptions-a-15-country-study/

Source 3: https://pubmed.ncbi.nlm.nih.gov/33684019/





Indicators for the "3Cs" Model¹



Note 1: Full indicator descriptions are available in the Appendix.

Note 2: Indicators for the confidence model are sourced from the Afrobarometer and are only available at the first administrative level.

Source: Malawi Demographic and Health Surveys, World Health Organization, Malaria Atlas Project "Global maps of travel time to health facilities"





Profiles of Vaccine Hesitant Segments

Fraym created each profile based on their unique combination of the 3Cs. The replicated Ipsos segments are not exhaustive combinations of the "3Cs" and do not add to the total population of adults in Malawi.

Segment Profile	Confident Enthusiasts	Enthusiastic Pragmatists	Vaccine Ambivalents	Vaccine Sceptics	COVID Cynics
Segment Description	Convinced of COVID threat and vaccine benefits. Would be quick adopters driven by social responsibility to protect their community.	Convinced of COVID threat and merits of a vaccine, but inhibited by practical barriers. Cost-benefit analysis of the process could cause uptake delay.	Not convinced of the threat of COVID as a disease and lack motivation to seek a vaccine, but few barriers to uptake. Could be moved by social norms and strong messaging.	Convinced of COVID threat, but scepticism around vaccine safety and efficacy inhibits perceived benefit and quick uptake.	Strongly hesitant of COVID threat and a COVID vaccine. Mistrust in the vaccine's purpose and advocates means they will be slow to vaccine adoption, if at all.
Level of Confidence	High	High	Moderate	Low	Low
Level of Convenience	High	Low	Moderate	Moderate	Moderate
Level of Complacency	Low	Low	High	Low	High
Potential Speed of Vaccine Uptake	Rapid	Delayed	Slow	Very Slow	Least Likely





How to use this analysis

Paired with WHO's 3Cs Framework, Fraym has mapped J&J's Consumer Segments to identify where vaccine uptake challenges are likely to occur.

Potential Use-Cases:

- Concentrate communication campaigns and media spending in specific geographic areas
- Target specific messages to niche audiences in prioritized geographies
- Optimize vaccine distribution
- Service & Product Delivery Planning

3 Levels of Views

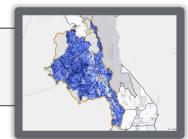
National

Patterns at the country level



District

Patterns across counties



Community

Patterns at the Sq. Km





Segment Mapping Use Cases

Mapping the 3Cs of Vaccine Hesitancy

Fraym created hyper-local maps of vaccine complacency and convenience and a state-level view of confidence (proxied by trust in government).

- Fraym used the 2016 Malawi Demographic and Health Survey to model vaccine complacency and convenience. To incorporate confidence, Fraym utilized state-level data from the 2020 Malawi Afrobarometer.
- Vaccine complacency indicators were created based on limited interactions with the health system as a proxy for low perceived risk for diseases and preventative actions. Vaccine convenience indicators focus on accessibility, as well as time and financial burdens associated with getting vaccinated. Fraym utilized multiple correspondence analysis (MCA) to create complacency and convenience indices and normalized index values from zero to one.¹
- Fraym created a confidence in government index using the 2020 Malawi Afrobarometer. The index averages trust in a range of public figures—including the Prime Minister, members of the House of Peoples' Representatives, the local government, and traditional or religious leaders—through trust towards these figures and disapproval of their performance. The state-level data was normalized from zero to one. All individuals living in that state receive the same score.

Note 1: To learn more about MCA, please visit https://www.sciencedirect.com/topics/computer-science/multiple-correspondence-analysis.

Note 2: The replicated Ipsos segments are not exhaustive combinations of the "3Cs" and do not add to the total population of adults in Malawi.



Demographics of Vaccine Segments

Fraym analyzed the demographic characteristics of each segment among adults aged 15-49.









	Confident Enthusiasts	Enthusiastic Pragmatists	Vaccine Ambivalents	Vaccine Sceptics	COVID Cynics		
Age							
15-34	63%	60%	47%	61%	49%		
35-49	37%	40%	53%	39%	51%		
		Education	on				
No education	12%	27%	14%	4%	2%		
Complete primary	25%	11%	19%	38%	34%		
Complete secondary	7%	1%	5%	7%	5%		
Higher education	1%	0%	0%	0%	0%		
		Regular Media Co	nsumption¹				
Newspaper/Magazine	8%	3%	8%	5%	10%		
Television	17%	2%	4%	10%	15%		
Radio	31%	24%	37%	28%	32%		
Asset Ownership							
Mobile phone	75%	28%	75%	84%	78%		
Television	21%	2%	5%	11%	15%		
Radio	53%	33%	48%	42%	54%		

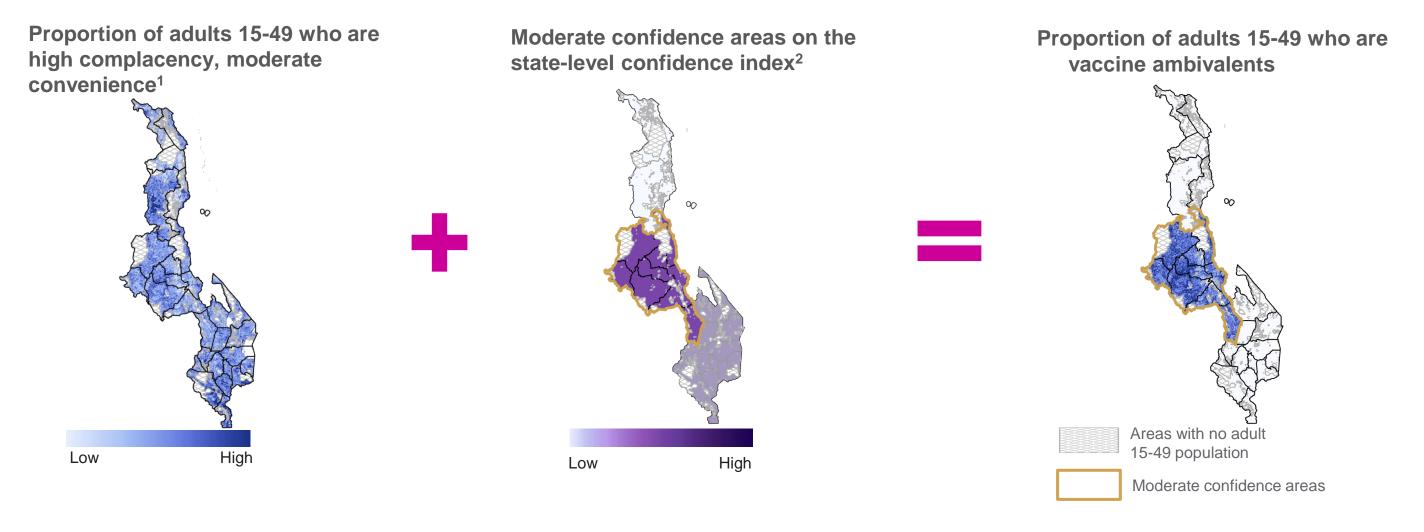
- Note 1: Regular media consumption is defined as listening to the radio, watching television, or reading a newspaper or magazine at least once per week.
- **Note 2:** Categories do not add to 100% due to response options not shown.





Identifying Vaccine Ambivalents

Vaccine ambivalents are not convinced of the threat of COVID as a disease and lack motivation to seek a vaccine, but face few barriers to uptake. This group could be moved by social norms and strong messaging.



Note 1: High complacency adults are adults who are in the third tercile of the complacency index, and moderate convenience in the second tercile of the convenience index.

Note 2: Moderate government confidence areas are those that fall into the second tercile of the state-level confidence index. Areas that are not moderate confidence were made transparent.

Source: 2016 Malawi DHS, 2020 Malawi Afrobarometer, Fraym





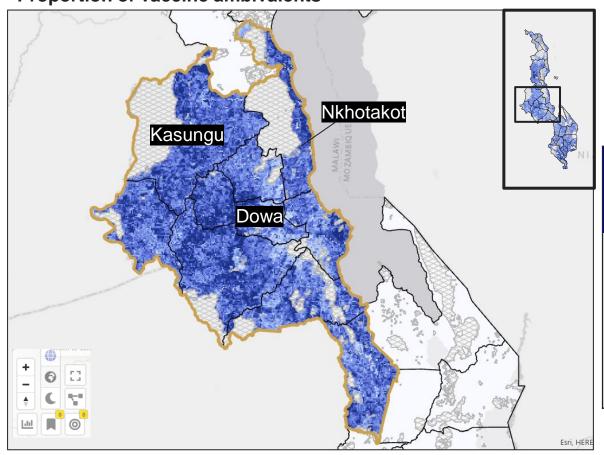
Top Districts: Vaccine Ambivalents



Tip: Use the DATA fraym® "Top Locations" tool to view locations with the highest values for your chosen indicator.

Fraym calculated the percentage of vaccine ambivalents and aggregated them to the district level to identify the areas most in need and least likely to receive vaccines.¹

Proportion of vaccine ambivalents



DATA fraym® allows users to target priority populations depending on specific programming needs. Fraym first identified the number of adults in a district, then found the proportion of high complacency adults. Of these high complacency adults, Fraym identified vaccine ambivalents who have moderate levels of convenience and confidence.

Rank (by % of vaccine ambivalents)	Region	District	Adult 15-49 population	High complacency (%)	Vaccine Ambivalents (%)	Vaccine Ambivalents (# of people)
1	Central	Dowa	404,605	34%	11%	43,699
2	Central	Kasungu	421,983	33%	10%	42,911
3	Central	Nkhotakota	179,667	34%	10%	18,156
4	Central	Mchinji	326,400	30%	10%	32,070
5	Central	Ntcheu	296,060	31%	10%	28,681



Note 1: The proportion of vaccine ambivalents was generated for every 1km² grid across Malawi and then aggregated to the district level.

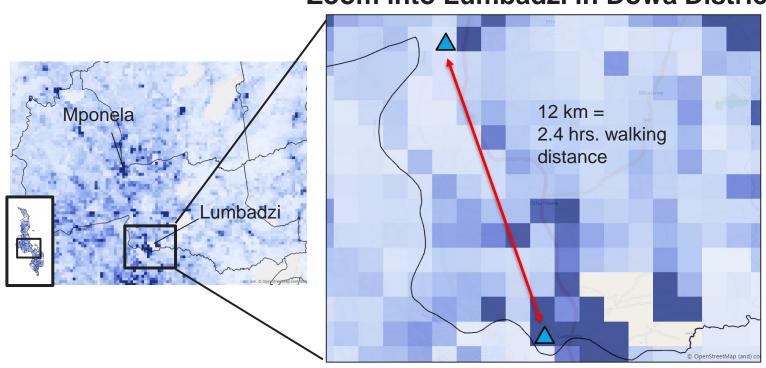




Vaccine Ambivalents: Hyperlocal View

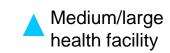
Communities with high numbers of vaccine ambivalent adults in Dowa district also have few medium and large health facilities. This means that access to vaccines, in addition to complacency, are important for this group.

Zoom into Lumbadzi in Dowa District



Number of adults who are vaccine ambivalent^{1,2}





Nearly 405,000 adults live in the **Dowa** district, of which **44,000 (11%)** are expected to be vaccine ambivalents.

- Hotspots of vaccine ambivalent adults in Dowa can be found in urban areas such as Mponela and Lumbadzi. This group may require a targeted outreach via mobile phone (75% ownership).
- Although there are health facilities present, they
 are relatively scattered in the district. In
 Lumbadzi, the only two medium-large facilities
 are 12 km apart. Dowa could benefit from pop-up
 clinics providing services to villages that are too
 far from established health clinics.

Note 1: Adults are defined as individuals aged 15-49.

Note 2: Map shows the number of adults who are high complacency and moderate convenience and are living in moderate government confidence areas.

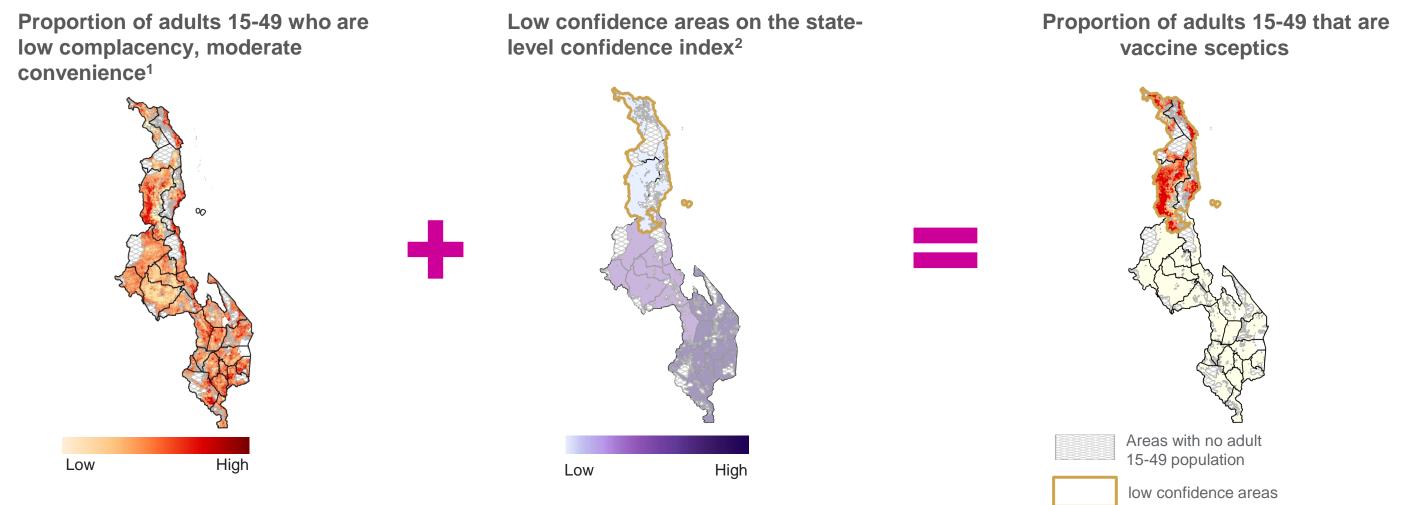
Source: 2016 Malawi DHS, 2020 Malawi Afrobarometer, Fraym





Identifying Vaccine Sceptics

Vaccine sceptics are convinced of COVID threat, but have scepticism around vaccine safety and efficacy inhibits perceived benefit and quick uptake. Trust in the vaccine and the system that delivers it are key levers for this segment.



Note 1: low complacency adults are adults who are in the first tercile of the complacency index, and moderate convenience in the second tercile of the convenience index. **Note 2:** low government confidence areas are those that fall into the first tercile of the state-level confidence index. Areas that are not low confidence were made transparent. **Source:** 2016 Malawi DHS, 2020 Malawi Afrobarometer, Fraym





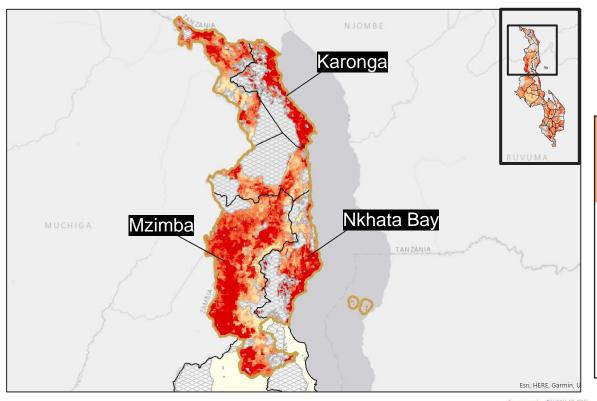
Top Districts: Vaccine Sceptics



Tip: Use the DATA fraym® "Top Locations" tool to view locations with the highest values for your chosen indicator.

Fraym calculated the percentage of vaccine sceptics and aggregated them to the district level to identify the areas most in need *and* least likely to receive vaccines.¹

Proportion of vaccine sceptics



DATA fraym® allows users to target priority populations depending on specific programming needs. Fraym first identified the number of adults in a district, then found the proportion of low complacency adults. Of these low complacency adults, Fraym identified vaccine sceptics who have moderate levels of convenience and low confidence.

Rank (by % of vaccine sceptics)	Region	District	Adult 15-49 population	Low complacency (%)	Vaccine Sceptics (%)	Vaccine Sceptics (# of people)
1	Northern	Karonga	170,582	32%	10%	55,225
2	Northern	Nkhata Bay	134,904	27%	9%	37,059
3	Northern	Mzimba	607,959	26%	8%	157,478
4	Northern	Chitipa	116,910	24%	7%	28,498
5	Northern	Rumphi	107,172	32%	6%	31,221

Note 1: The proportion of vaccine sceptics was generated for every 1km² grid across Malawi and then aggregated to the district level.



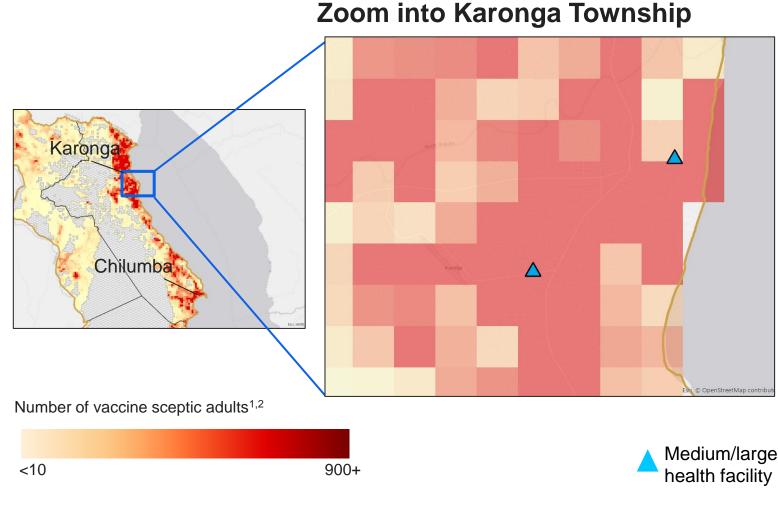
Low



High

Vaccine Sceptics: Hyperlocal View

Vaccine sceptics are convinced of the COVID thread, but low confidence in the government may make vaccine outreach more challenging.



171,000 adults live in the Karonga district, of which 55,000 (10%) are expected to be vaccine sceptics.

- Most vaccine sceptics live in towns such as Karonga Township and Chilumba. This group may take the vaccine at a slower rate given their moderate convenience and low confidence.
- Low confidence in the government poses a significant challenge. However, given the high mobile phone ownership (84%), a media campaign via text may be viable.

Note 1: Adults are defined as individuals aged 15-49.

Note 2: Map shows the number of adults who are low complacency and moderate convenience and are living in low government confidence areas.

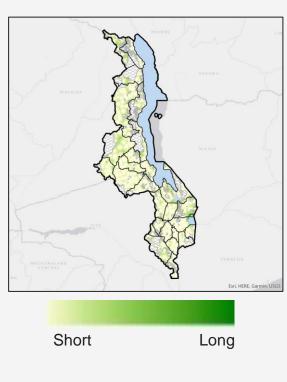
Source: 2016 Malawi DHS, 2020 Malawi Afrobarometer, Fraym



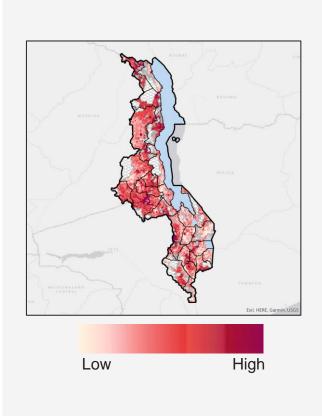
Community-Level Context

Contextual information at the community level can help inform broad RCCE efforts. All of these data indicators are in the DATAfraym® dashboard with geographic precision at the 1km², district, and region levels.

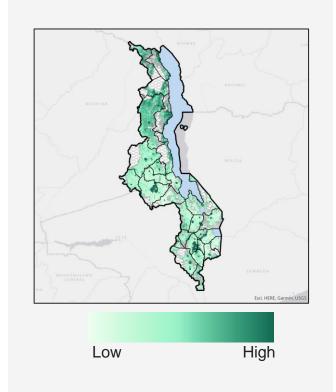
Walking Distance to Health Facility (min)



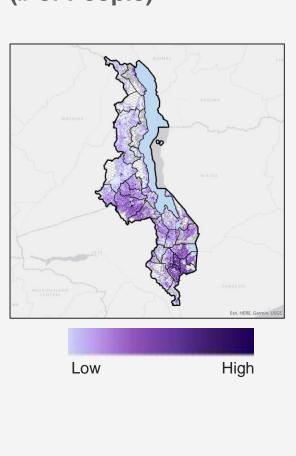
Radio Listenership (%)



Mobile Phone (%)



Elderly Population 60+ (# of People)



Source: 2016 Malawi DHS, 2020 Malaria Atlas Project





Data and Methods

Indicators for COVID-19 Vaccine Complacency

Fraym modeled COVID-19 vaccine complacency based low interactions with the health system.

Indicator	Description				
Interactions with the health system:	Interactions with the health system: Individuals living in households with limited preventative or birth-related services.				
No previous childhood vaccination	Individual lives in a household where at least one under 5 child has not been vaccinated for polio, diphtheria, pertussis (whooping cough), and tetanus (DPT), hepatitis B, or measles.				
Post-natal care	Individual lives in a household where at least one woman received a postnatal check within 2 months of giving any births in the past 0-35 months.				
Family planning	Individual lives in a household where at least woman has heard of family planning at a health facility in the past 12 months.				
Place of delivery	Individual lives in a household where a woman has given birth at a government hospital, government health center, government health post, NGO health facility, or a private hospital/clinic for any births in the past 0-35 months.				

Source: 2016 Malawi DHS

Note 1: Fraym removed female visits to health facilities or visits by fieldworkers from the vaccine complacency model due to lack of variation.



Indicators for COVID-19 Vaccine Convenience

Fraym modeled COVID-19 vaccine convenience with accessibility, time burden, and financial burden.

Indicator	Description		
Accessibility: Geographic distance will determine that can be expected	how physically feasible it is to receive a vaccine. The type of health facility will determine service capacity as well as the quality of service		
Distance to health facility (walking)	Walking time to nearest health facility using least cost distance. Travel time will determine how physically feasible it is to receive a vaccine.		
Distance to health facility (driving)	Driving time to nearest health facility using least cost distance. Travel time will determine how physically feasible it is to receive a vaccine.		
Car, truck, or scooter ownership	Individual lives in a household that owns a car, truck, or scooter.		
Access to a mobile phone	Individual lives in a household that owns a mobile phone. If vaccine appointments are booked primarily using a smartphone or internet connection, lack of the necessary technology may make it more difficult book an appointment and receive a jab.		
Time burden: Concerns about time, cost or vaccin	e site accessibility may deter vaccine uptake.		
Caregiving/domestic responsibilities	Individual lives in a household with a child under 5 or an adult over 60.		
Time to fetch water	Individual lives in a household that must travel longer than 1 hour to fetch drinking water.		
Financial burden: Lower financial resources may affect the ability to receive a vaccine, particularly if the vaccine is perceived as costly.			
Bottom wealth quintile	Individual lives in a household that is in the bottom of the DHS wealth quintile.1		
Unemployment or agricultural employment	Individual is employed in the agriculture sector or is unemployed.		

Source: 2016 Malawi DHS, 2020 Malaria Atlas Project "Global maps of travel time to healthcare facilities"

Note 1: The wealth index is a composite measure of a household's cumulative living standard from the DHS survey, calculated using information on household asset ownership, housing materials, and access to water and sanitation services. The first quintile is the poorest while the fifth quintile is the wealthiest.



Indicators for Confidence in Government

Fraym created a trust in government index using state-level Afrobarometer data.

Indicator	Description
Trust towards public figures	
Trust in public institutions	Individual trusts the Prime Minster, House of Peoples' Representatives, the local government, ruling party, opposition party, traditional leaders, or religious leaders "not at all" or "just a little".
Disapproval of public figures' performance	
Disapproval of performance	Individual "strongly disapproves" or "disapproves" of the way the Prime Minister, their member of the House of Peoples' Representatives, their elected local government leader, or traditional leaders have performed their jobs over the past 12 months.

Source: 2020 Malawi Afrobarometer



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ACQUIRE DATA

Geo-tagged household surveys

Satellite imagery

Partner datasets

Mobility data from network operators



Validate Clean Geospatially enable



MACHINE LEARNING

Proprietary algorithms
Human-centric QA/QC
Automation



GEOSPATIAL INSIGHT

API enabled

Analytic services

Front-end tools

Thank you.

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