

# STATISTICAL ANALYSIS PLAN



## ACTION LEVERAGING EVIDENCE TO REDUCE PERINATAL MORTALITY AND MORBIDITY IN SUB-SAHARAN AFRICA (ALERT) TRIAL

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### VISION HISTORY

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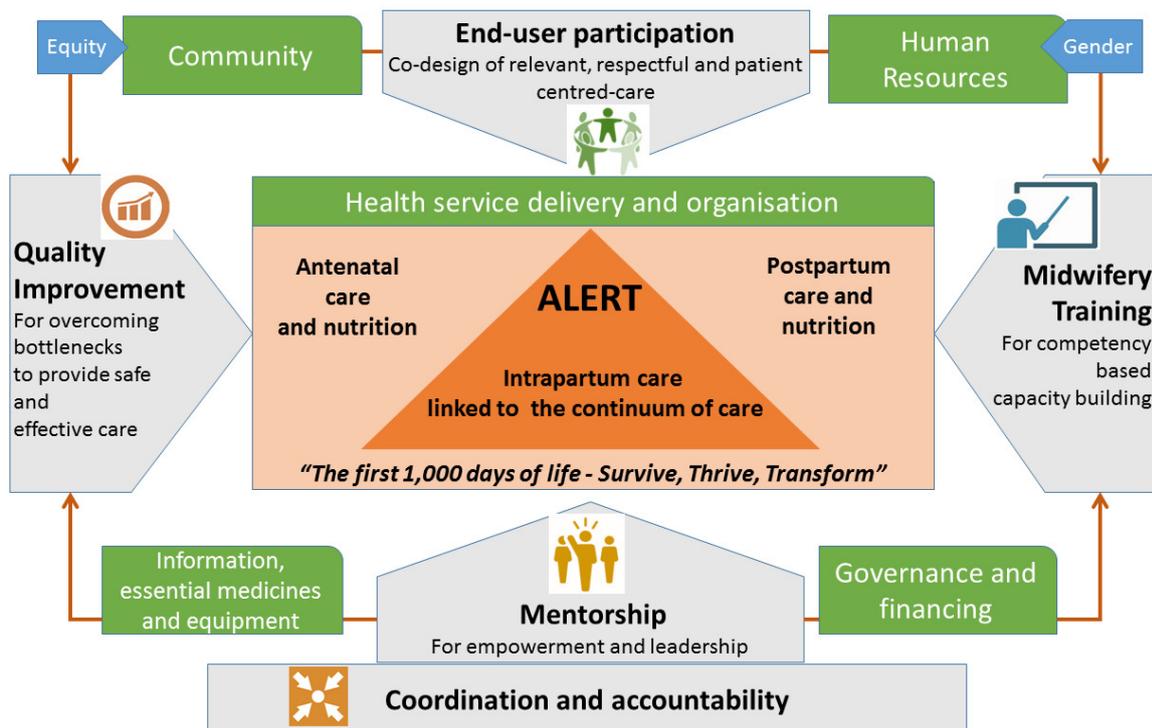
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# 1. STUDY BACKGROUND

The Action Leveraging Evidence to reduce perinatal Mortality and morbidity in sub-Saharan Africa (ALERT) trial was set to develop and evaluate a comprehensive and multilevel intervention covering intrapartum care and midwifery. The ALERT trial and interventions targeted hospital maternity units and included i) end-user participation of women, families, and midwifery providers to co-design the intervention; ii) in-service midwifery competency-based training; iii) empowerment and leadership mentoring of maternity unit leaders, and iv) quality improvement in the maternity ward, supported by district-based bi-annual coordination and accountability meetings (Fig. 1).

Figure 1: ALERT Conceptual Framework



## 2. STUDY OBJECTIVES

### 2.1 Specific Objective

The ALERT trial had three specific study objectives

1. To assess the ALERT intervention's impact in hospital maternity units on perinatal and maternal health outcomes, including women's experience of care.
2. To evaluate the process of implementation of the intervention to understand what works for whom and under what situation.
3. To conduct a cost-effectiveness analysis of the ALERT intervention.

### 2.2 Assessment of Objectives

A summary of our approach to answering the objectives is given in Table 1. In this Analysis Plan, we give details for only objective one (details about other study objectives are given elsewhere).

Table 1: Methods and data sources by study objective

Objective	Methods	Data Source	Notes and links
<p>Obj 1. To assess the ALERT intervention's impact in hospital maternity units on perinatal and maternal health outcomes, including women's experience of care.</p>	<p>Trial profile</p> <p>Assessment of recruitment rate, loss to follow-up and records completeness/incompleteness</p> <ul style="list-style-type: none"> <li>- Number of records entered the REDCap database in comparison to expected number of records</li> <li>- Number of incomplete records</li> </ul> <p>Descriptive statistics (counts, proportions, mean, median, SD, max, min)</p> <ul style="list-style-type: none"> <li>- Study outcomes split by intervention arm</li> <li>- Additionally, we will summarise the primary outcome split by both site and by intervention condition.</li> </ul> <p>Regression models fitted for the primary outcome</p> <ul style="list-style-type: none"> <li>- Generalised Estimation Equations models will be used to adjust for the effect of clustering and repeated measures over time.</li> <li>- Intra cluster correlation (ICC) coefficients will be re-evaluated using the study data.</li> <li>- Unidirectional crossover as well as any potential secular trends may be incorporated with a time effect</li> <li>- Sensitivity analysis</li> </ul>	<p>Trial data</p> <p>Exit interviews</p>	<p>Our primary outcome is in-facility early perinatal mortality defined as in-facility (fresh) stillbirth and 24-hour neonatal mortality</p>

Objective	Methods	Data Source	Notes and links
	Women's experiences of care <ul style="list-style-type: none"> <li>- Descriptive statistics for each hospital, country (means and proportions)</li> <li>- Distribution of the of optimal care and number of interventions received.</li> <li>- Regression models to predict factors that determine optimal women's care</li> <li>- Spatial analysis? – Health seeking behaviours vis-a-viz care seeking</li> </ul>		
Obj 2. To evaluate the process of implementation of the intervention to understand what works for whom and under what situation.	Trial profile  Descriptive statistics (counts, proportions, mean, median, SD, max, min)	WP7 – [WP7 should have a protocol]	
Obj 3. To conduct a cost-effectiveness analysis of the ALERT intervention.	Cost effectiveness analysis <ul style="list-style-type: none"> <li>- Provider's perspectives</li> <li>- Effects per outcomes</li> <li>- Incremental cost-effectiveness ratio (ICER)</li> <li>- Sensitivity analysis</li> </ul>	Costing data	

## 3. STUDY DESIGN

### 3.1 Overview

A stepped-wedge cluster-randomized design with a nested process evaluation based on realist evaluation was employed in the ALERT trial. This design was selected as the interventions were administered at the hospital level, and a stepped-wedge design was chosen to mirror scale-up for policy buy-in and for statistical efficiency as larger cluster-level differences are anticipated.

### 3.2 Sample Size

The ALERT trial was carried out in 16 hospitals, with four hospitals selected in each of the four countries: Benin, Malawi, Tanzania, and Uganda. The selection criteria for the hospitals required them to have a minimum of 2500 births per year, which provided sufficient statistical power (75%-80%) to detect a 25% reduction in the in-facility early perinatal mortality rates with a 95% confidence interval. This also allowed for the assessment of other secondary outcomes such as maternal morbidity, hypoxic-ischemic insults, neonatal seizures, and more (see Figure 2).

Figure 2: Sample size and Power calculations for ALERT trial outcomes

	<b>Baseline estimates</b>	<b>expected change</b>	<b>expected value</b>	<b>ICC</b>	<b>Power</b>
<b>Perinatal e-registry (Cluster size 1500 births)</b>					
Stillbirth	1.0%	30%	0.7%	0.0024	83%
Stillbirth	1.0%	30%	0.7%	0.024	82%
Perinatal mortality	2.0%	25%	1.5%	0.0024	93%
Perinatal mortality	2.0%	25%	1.5%	0.024	92%
Perinatal mortality	1.4%	25%	1.1%	0.0024	70%
Hypoxic-ischaemic insults	4.0%	30%	2.8%	0.0024	100%
Hypoxic-ischaemic insults (sub-group)	4.0%	30%	2.8%	0.0024	79%
Neonatal seizures	0.5%	50%	0.3%	0.0024	78%
Cesarean section	15.0%	20%	18.0%	0.01	100%
Maternal morbidity	3.0%	30%	2.1%	0.5	99%
<b>Interviews post-partum (Cluster size 50 interviews)</b>					
Responsiveness	70.0%	15%	80.0%	0.1	97%
Mistreatment	4.0%	75%	1.0%	0.1	77%
Breastfeeding	60.0%	20%	71.0%	0.1	98%

### 3.3 Randomization and Blinding

The ALERT trial employed randomization, which was carried out by an independent statistician after obtaining consent from the participating hospitals. Stratification by country was used to enable random selection of four hospitals in each country (refer to randomization syntax in Figure 3a). Participants (hospitals) will not be blinded to the intervention, as with all training and quality improvement interventions. However, the women and their families receiving care at these hospitals will not be aware of the exact implementation stage of ALERT. The final sequence of ALERT intervention and data collection is illustrated in Figure 3b.

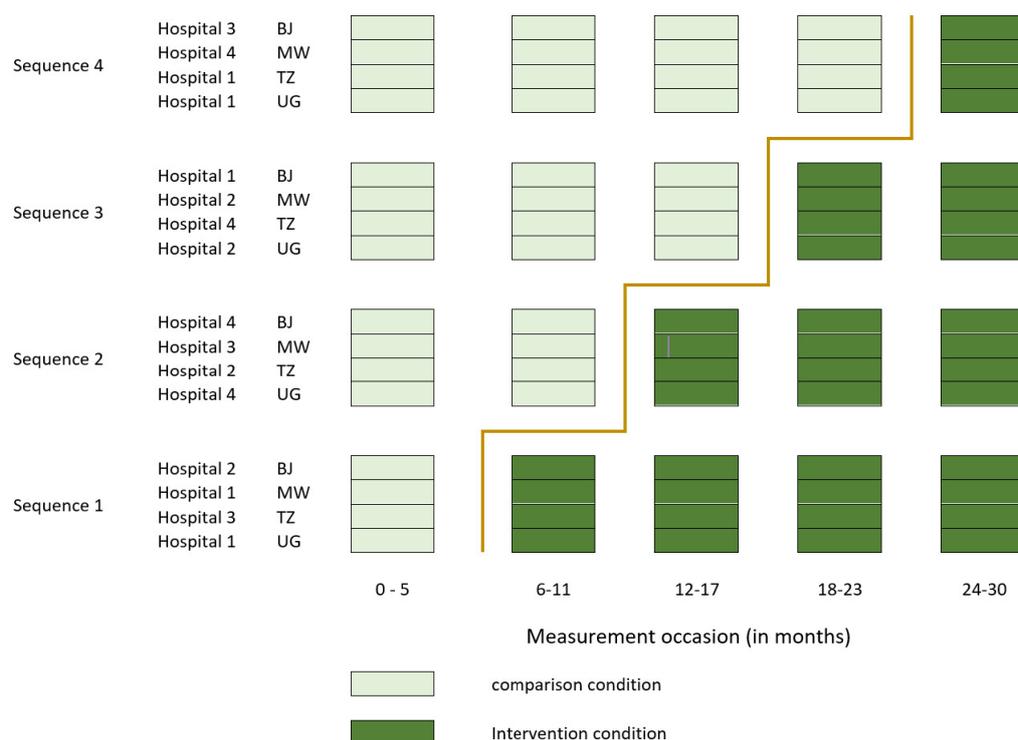
Figure 3a: Randomisation syntax file

```

1  /*
2
3  Objective: This do-file is showing how to create a file with the randomized sequence
4  Date: 20200617
5  Author: Orsini N
6  */
7
8  * cd "/Users/nicors/OneDrive - KI.SE/Mac/Research/claudia/"
9
10 /*
11 * Import into Stata all these data contained in the Excel file
12
13 local list = "Benin Malawi Tanzania Uganda"
14 foreach s of local list {
15     import excel "Hospitals for randomization .xlsx", sheet("`s'") clear
16     gen state = "`s'"
17     save `s', replace
18 }
19
20 * I append all these 4 stata dataset into a single dataset
21
22 use Benin, clear
23 append using Malawi Tanzania Uganda
24
25 rename B name
26 drop A
27 bysort state: gen code = _n
28 order state code
29 list , sepby(state) clean noobs
30
31 /*
32
33 * Another approach. Input the data manually. As shown below
34
35 clear
36 input str10 state code str50 name
37     Benin      1      "Centre National Hospitalier Hubert Maga"
38     Benin      2      "Hopital de Zone Come-Grand Popo Houeyogbe Bopa"
39     Benin      3      "Policlinique Bon Samaritan"
40     Benin      4      "Centre Hospitalier et Dépt Zou-Collines"
41     Malawi     1      "Mchinji Hospital"
42     Malawi     2      "Mitundu Rural Hospital"
43     Malawi     3      "Ntcheu D Hospital"
44     Malawi     4      "St Gabriels Hospital"
45     Tanzania  1      "Newala district Hospital"
46     Tanzania  2      "Mkomaindo Hospital"
47     Tanzania  3      "Ndanda Hospital"
48     Tanzania  4      "Nachingwea Hospital"
49     Uganda    1      "Bugiri General Hospital"
50     Uganda    2      "Iganga Referral Hospital"
51     Uganda    3      "Jinja Regional Hospital"
52     Uganda    4      "Kamuli Mission Hospitla"
53 end
54
55 * ssc install shuffle
56 * shuffle 1 2 3 4, num noisily
57
58 gen rand = .
59
60 * Set the seed to reproduce the randomization
61
62 set seed 20200617
63
64 local list = "Benin Malawi Tanzania Uganda"
65 qui foreach s of local list {
66     local j = 1
67     shuffle 1 2 3 4, num noisily
68     local seq = r(list)
69     foreach r of local seq {
70         replace rand = `r' if code == `j' & state == "`s'"
71         local j = `j' + 1
72     }
73 }
74
75 list, sepby(state)
76
77 export excel using "results_rand_200617", firstrow(variables) replace
78
79 exit

```

Figure 3b: Sequence of the ALERT Trial intervention



### 3.4 Study Assessments

To achieve objective one, we will conduct two assessments. Firstly, we will abstract trial data from hospital registries and enter it into a REDCap application. Secondly, we will conduct exit interviews with hospital clients to capture and evaluate their care-related experiences while in the hospitals.

## 4. STUDY POPULATIONS

### 4.1 Subject Disposition

Subject disposition in the ALERT trial refers to the movement of participants throughout the study, from screening to the completion of the intervention and follow-up. In the ALERT trial, subject disposition will be tracked for both hospital-level and patient-level participants.

At the hospital level, subject disposition will involve tracking the participation of the hospitals throughout the study. This includes monitoring their consent to participate, enrolment in the study, completion of the intervention, and follow-up data collection. Any dropout or non-compliance of the hospitals will also be recorded and reported.

At the patient level, subject disposition will involve tracking the participation of individual patients who are receiving care at the participating hospitals. This includes monitoring their enrolment in the study, completion of the intervention, and follow-up data collection. Any dropout or loss to follow-up of the patients will also be recorded and reported.

Subject disposition is important in clinical trials as it provides information on the success of the study in terms of recruitment and retention. It also allows for the assessment of potential biases and the

generalizability of the results. Therefore, in the ALERT trial, careful monitoring of subject disposition will be conducted to ensure the integrity and validity of the study findings.

## 4.2 Definition of Populations for Analysis

To ensure that the results of the ALERT trial are relevant to its intended audience, different populations will be used for analysis. The primary analysis will involve all women who gave birth in the 16 participating hospitals, regardless of whether they received the intervention or not. This will be used to assess the impact of the ALERT intervention on early perinatal mortality rates, the primary outcome of the trial.

The secondary analysis will focus on women who received the intervention during their hospital stay. This population will be used to assess the impact of the intervention on other outcomes such as maternal morbidity, hypoxic-ischaemic insults, and neonatal seizures. Additionally, subgroups based on specific covariates, such as birth weight, mode of delivery, time of delivery, and type of outcome, may be analysed to further explore the impact of the intervention on specific populations of interest.

By using multiple populations for analysis, the ALERT trial will provide a comprehensive understanding of the impact of the intervention on different outcomes and populations. This approach ensures that the results are applicable to a broad range of stakeholders in the field of maternal and child health.

## 4.3 Definition of Sub-Group Population in Different Analyses

In the ALERT trial, sub-group populations are defined based on specific covariates, such as birth weight, mode of delivery, time of delivery, and type of outcome, in different analyses. These sub-groups are used to investigate the impact of the ALERT intervention on specific populations of interest and to identify potential effect modifiers.

For example, in the primary analysis, the sub-group population will consist of all women who gave birth in the 16 participating hospitals during the study period, regardless of whether they received the ALERT intervention. This population will be used to assess the impact of the intervention on early perinatal mortality rates.

In the secondary analysis, the sub-group population will include all women who received the ALERT intervention during their hospital stay. This population will be used to evaluate the impact of the intervention on secondary outcomes such as maternal morbidity, hypoxic-ischaemic insults, and neonatal seizures.

Additional sub-group analyses may be conducted based on specific covariates of interest, such as birth weight or mode of delivery. These sub-groups will allow for a more detailed examination of the impact of the intervention on specific populations of interest and to identify potential effect modifiers that may impact the efficacy of the ALERT intervention.

Overall, the use of sub-group populations in different analyses will provide a more nuanced understanding of the impact of the ALERT intervention on different populations and outcomes, which can inform the development of targeted interventions for improved maternal and child health outcomes.

#### 4.4 Efficacy Evaluable in the ALERT trial

In the ALERT trial, the efficacy evaluable population will be defined as all women/hospitals who received the ALERT intervention and completed the study as per protocol. This population will be used to evaluate the efficacy of the intervention in improving perinatal outcomes.

To be considered as efficacy evaluable, participants must have received the intervention in full, without any major deviations from the study protocol, and completed the study follow-up assessments. This approach will ensure that the population analysed reflects the impact of the intervention as intended.

The efficacy evaluable population will be used in the secondary analysis of the trial, which will evaluate the impact of the ALERT intervention on outcomes such as maternal morbidity, hypoxic-ischaemic insults, and neonatal seizures. This analysis will provide important information on the potential benefits of the intervention in improving maternal and neonatal health.

Overall, the efficacy evaluable population in the ALERT trial is critical in determining the effectiveness of the intervention and ensuring that the results of the trial are reliable and generalizable to the intended population.

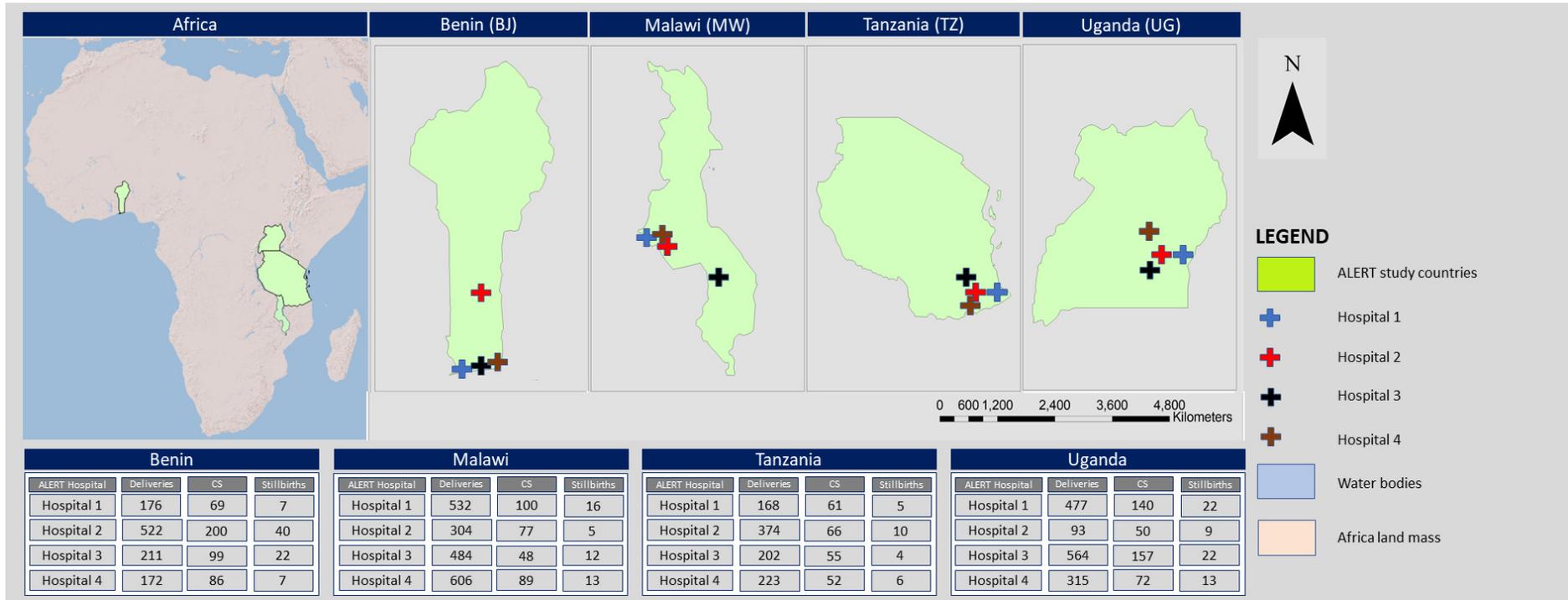
#### 4.5 Intent-to-treat (ITT)

The intent-to-treat (ITT) analysis is a statistical method used in clinical trials that evaluates the effectiveness of an intervention in real-world conditions by analyzing all randomized participants, regardless of their adherence to the protocol or treatment received. In the ALERT trial, the ITT analysis will be used to assess the impact of the ALERT intervention compared to standard care. This method provides a more realistic estimate of the intervention's effectiveness in a real-world setting, but may underestimate the treatment effect as it includes all randomized participants, including those who did not receive the intended treatment or dropped out of the study.

#### 4.6. Study Sites and Populations

The ALERT trial, which began in April 2021, purposefully selected four hospitals in each country through consultations with the respective country's Ministry of Health (refer to Figure 4). All selected hospitals met the following criteria: (i) a minimum caseload of 2,500 births per year, based on trial sample size calculation; (ii) availability of caesarean section and blood transfusion services; (iii) preferably located in rural districts; and (iv) consisting of a mix of typical public but also private-not-for-profit (faith-based) hospitals. The inclusion of public and private-not-for-profit hospitals aimed to reflect the typical hospital landscape in sub-Saharan Africa and improve the generalizability of our results.

Figure 4: Map showing the ALERT trial countries and key indicators for the study hospitals



## 4.7 Major Protocol Deviations

In any clinical trial, it is essential to adhere to the study protocol to ensure the validity and reliability of the results. However, sometimes deviations from the protocol may occur due to unforeseen circumstances or human error. In the ALERT trial, major protocol deviations will be defined as any deviation that affects the primary outcome or the safety of the participants.

Examples of major protocol deviations may include but are not limited to the following:

- Administration of the ALERT intervention to a non-intervention hospital
- Failure to administer the ALERT intervention in an intervention hospital
- Early termination of the intervention period
- Failure to obtain informed consent from participants
- Major errors in data collection or analysis

Any major protocol deviations will be reported to the trial steering committee, and appropriate action will be taken to address the deviation and mitigate its impact on the trial results. The impact of major protocol deviations on the primary outcome and safety of the participants will be assessed through sensitivity analyses.

## 5. STATISTICAL ANALYSIS

### 5.1. Outcome measures

The outcomes of the ALERT trial will be measured for all admitted mothers who deliver in the study hospitals and their pregnancy outcomes (live births, stillbirths, neonatal deaths) that have recorded in the e-registry since March 2021 when the first hospitals

### 5.2. Analytical approaches

The Makerere University School of Public Health and Karolinska Institutet, in collaboration with the London School of Hygiene & Tropical Medicine, will conduct statistical analysis of primary and secondary outcomes using Stata Standard Edition 17 and R statistical and programming software. The analysis will follow an "intention-to-treat" approach, comparing ALERT intervention clusters (hospital maternity wards) with comparison clusters where care is provided according to national standards.

#### *Descriptive statistics*

We will report the number of records, completed records, admissions, and status, as well as mode of delivery and birth outcomes by hospital per time period (refer to dummy table 2).

In this study, we will employ interrupted time series analysis to assess trends in the perinatal registry data collected over a 30-month period. We will present maternal characteristics and outcomes for each hospital, grouped according to exposure (control vs ALERT intervention). While we anticipate no differences in the demographic characteristics of mothers admitted to each hospital during the study period, we expect changes in the number of admissions and outcomes recorded at each hospital as a result of the intervention. We hypothesize that the ALERT interventions will lead to a reduction in adverse pregnancy outcomes, as they are intended to enhance service provision and data quality at the hospitals.

As each hospital operates within a unique population mix, differences may arise within the hospital depending on when they were randomized to the intervention. To account for potential imbalances across hospitals and sequences, we will adjust demographic covariates, although we may not test for demographic differences between the control and intervention groups.

We will present categorical attributes as frequency and percentage, and continuous attributes as mean and standard deviation or median and interquartile range, based on their distribution (see dummy table 3). Additionally, we will present primary outcomes graphically by month (see dummy figure 5), and secondary outcomes will be presented as tables or figures.

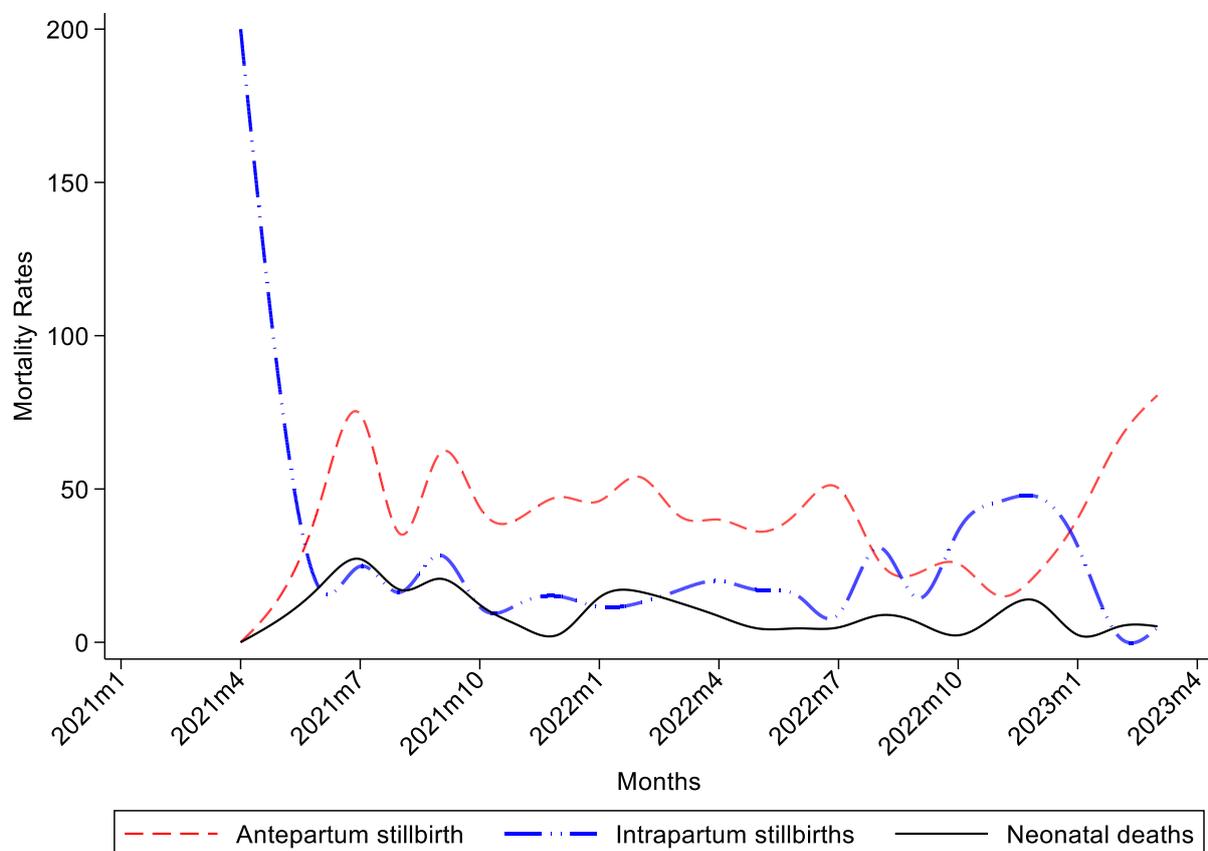
Table 2: Example of how number of admissions and data completeness could be presented

Hospital	Total number of records	No. completed records	Referral status and state at admission				Mode of delivery							Birth outcomes			
			Not referred	Yes – not in labour	Yes – in labour	Yes – after birth	SVD	C/S	Vacuum	Forceps	Breech	Singleton	Multiple births	Live birth	Fresh stillbirth	Macerated stillbirth	Neonatal deaths
Benin Hospital 1																	
Sequence 1																	
Sequence 2																	
Sequence 3																	
Sequence 4																	
Hospital 2																	
Sequence 1																	
Sequence 2																	
Sequence 3																	
Sequence 4																	
Etc.																	
Malawi Hospital 1																	
Sequence 1																	
Sequence 2																	
Sequence 3																	
Sequence 4																	
Hospital 2																	
Sequence 1																	
Sequence 2																	
Sequence 3																	
Sequence 4																	
Etc.																	

Table 2

Hospital	Variable	Control Arm	Treatment Arm
Benin			
Hospital 1	<ul style="list-style-type: none"> <li>• Number of records</li> <li>• Age</li> <li>• Status at admission</li> <li>• 1<sup>st</sup> Stage of labour mgmt</li> <li>• Mode of delivery</li> <li>• PNC checks</li> <li>• Initiation of breast feeding</li> <li>• Length of stay in hospital</li> <li>• N stillbirths</li> <li>• N livebirths</li> <li>• N neonatal deaths</li> <li>Etc.</li> </ul>		
Hospital 2			
Hospital 3			
Hospital 4			

Figure 5 – Example of graphical presentation of of monthly mortality rates for Jinja regional referral hospital



### *Multivariable analysis*

The study will include an analysis of seasonal variations in birth weight and neonatal mortality rates. While stillbirths and early neonatal mortality have declined slowly in the past, we anticipate annual declines of at least 2%. We will examine secular trends across different strata, including countries and hospitals, and estimate the degree of variation in effects across clusters, as recommended by Hemming et al.

GEE is a statistical method that can be used to account for correlated data, such as repeated measures or clustered data. In the case of the ALERT trial, GEE may be a useful approach to account for the correlation between outcomes within the same hospital and to adjust for potential time-trends or other cluster-level factors that may impact the outcomes.

One potential challenge in the use of GEE is the limited number of clusters and sequences in the ALERT trial. To address this, the analysis will initially focus on within-period and within-cluster comparisons to account for the small sample size and clustering. This may involve comparing outcomes within the same hospital and period, as well as conducting subgroup analyses based on relevant covariates.

Once the within-cluster analyses have been completed, GEE can be used to model the relationship between the intervention and outcomes while adjusting for clustering, time-trends, and other relevant factors. This approach may improve the precision of the estimates and help to account for potential confounding factors, such as changes in hospital policies or staffing over time.

### *Model building using GEE -*

The estimate of the mortality risk for the ALERT intervention effect (control vs intervention arms) with their 95% confidence intervals shall be presented for all models fitted.

Models shall be fitted using the following numerical notations

*I* clusters ( $i=1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16$ )  
*M* time periods ( $j=1, 2, 3, 4, 5$ )  
*N* outcomes i.e., stillbirths ( $k=1, 2, \dots, N$ ), sampled per cluster per time point (cross-sectional cohort)  
Treatment indicator ( $T_{ij}$ ), equals 1 if intervention present at cluster *I* at time *J*, else it is 0  
A fixed treatment effect ( $\vartheta$ )  
Fixed time effect ( $\gamma_j$ ) (one parameter if calendar time used as continuous variable, otherwise vector)  
Fixed effects [ $\beta$ ] for patient-level demographics  
Patient-level adjustment variables [ $X_k$ ]  
Random cluster effect ( $\alpha_i$ )  
Residual noise ( $\epsilon_{ijk}$ )

Model 1 – Shall be fitted with an unadjusted before and after analysis of the effect of the intervention. For this model, the time effect shall be ignored.

$$\text{Logit}(Y_{ik}) = \vartheta * T_i + \alpha_i + \epsilon_{ik} \dots \dots \dots (1)$$

Where  $Y_{ik}$  denotes the probability of the outcome for  $k^{th}$  individual in the  $i^{th}$  cluster in the  $m^{th}$  period

$\vartheta$  denotes the log odds for the treatment/intervention variable (1= treatment, 0=control) in  $j^{th}$  cluster

Model 2 – Shall be fitted with adjusted time period (step) to enable us to examine whether any potential ALERT intervention effect is related only to the intervention or also to an independent effect of calendar time

$$\text{Logit}(Y_{ijk}) = \vartheta * T_{ij} + \gamma_j + \alpha_i + \epsilon_{ijk} \dots \dots \dots (2)$$

Where  $\gamma$  denotes the log odds for the effect of Time (vector if effect not linear)

The use of calendar time as a variable in the analysis poses a risk of confounding, as there may be other factors or events (such as changes in hospital practices or updates to Ministry of Health guidelines) that could impact the outcome measure in both the control and exposed arms.

The impact of this phenomenon may vary from being non-existent or gradual (slowly developing trend) to sudden (rapid and immediate adoption of a new approach that produces a powerful effect). To account for these variations, the model will initially incorporate time as a categorical variable, followed by a linear variable, and the most suitable fitting will be selected accordingly.

Model 3 - Shall be fitted with adjusted patient-level attributes

$$\text{Logit}(Y_{ijk}) = \vartheta * T_{ij} + \gamma_j + [\beta]X_{ijk} + \alpha_i + \epsilon_{ijk} \dots \dots \dots (3)$$

Where  $\beta$  denotes the log odds for matrix of X covariates for the  $k^{th}$  episode in the  $i^{th}$  hospital and the  $j^{th}$  time.

In the statistical analysis, we plan to adjust for potential confounding factors by including covariates such as age at hospitalisation (which may be treated as a continuous variable or grouped by age-range as appropriate), gestational age at admission or birth, previous antenatal care history, mode of delivery, time of delivery, referral status, among others. This adjustment will help us to isolate the effect of the ALERT intervention from other factors that could influence maternal and neonatal outcomes. We will also explore potential effect modification by subgroup analyses based on relevant characteristics, such as maternal age and parity.

Model 4 - Model 4 will incorporate time as a treatment effect modifier to assess how the impact of the ALERT intervention changes over time. This will enable us to determine the time required to observe a full-size effect of the intervention on the primary outcome and whether the size of the effect is maintained over time. We will fit the model using interrupted time series analysis, with time as a continuous variable measured as time since exposure to the intervention. In addition, we will include covariates to adjust for potential confounding factors that may affect the outcome.

$$\text{Logit}(Y_{ijk}) = \vartheta * T_{ij} + \gamma_j + [\beta]X_{ijk} + \omega Q_{ij} + \alpha_i + \epsilon_{ijk} \dots \dots \dots (4)$$

This equation 4 represents a model for examining the interaction between time and treatment (variable Q) on the log odds of the outcome for a given hospital (I) at a specific time (J). The variable Q will be analyzed as a numerical variable, where 0 represents any control period and 1, 2, etc. represent the first, second, and subsequent exposure steps of the intervention. The variable  $\omega$  represents the log odds of this interaction, and it will be fitted as both a continuous and categorical

variable to determine the most appropriate fitting. This model will help us understand how the impact of the intervention changes over time and whether the effect size is maintained over time

### ***Sub-group analysis***

Sub-group analysis will be conducted to investigate the effect of the intervention on specific sub-groups of interest, such as stratification by birth weight, mode of delivery, time of delivery, and type of outcomes. The decision to conduct sub-group analysis will depend on the power and sample size of the study, as well as the plausibility of the sub-groups being important for clinical or policy decision-making. These sub-group analyses will be conducted using appropriate statistical methods and will be interpreted cautiously, considering the potential for chance findings and the need for replication in future studies.

### **5.3 Pooling of Sites**

Pooling of sites refers to the process of combining data from multiple study sites or clusters to increase the power and generalizability of the results. In the ALERT trial, pooling of sites will be used to increase the statistical power of the study and to obtain more precise estimates of treatment effects.

Pooling of sites will involve combining data from the 16 participating hospitals to create a larger sample size. This will improve the precision of the estimated treatment effect, reduce the likelihood of type II errors (false negative results), and increase the generalizability of the findings to a broader population of women and infants.

However, pooling of sites can also introduce heterogeneity across sites, which can affect the internal validity of the study. To address this, the ALERT trial will use appropriate statistical methods, such as random effects models, to account for the potential heterogeneity between sites.

In summary, pooling of sites in the ALERT trial will provide several benefits, including increased power and generalizability of the results. However, it is important to carefully consider and account for any potential heterogeneity between sites in the analysis to ensure the validity of the study results.

### **5.4 Interim Analyses**

Interim analyses are pre-planned statistical evaluations of accumulating data during a clinical trial to assess the efficacy and safety of the intervention. In the ALERT trial, interim analyses will be conducted to evaluate the safety and efficacy of the ALERT intervention at predetermined time points. The interim analyses will be conducted by an independent Data and Safety Monitoring Board (DSMB), which is an independent group of experts responsible for monitoring the safety and efficacy of the trial.

The DSMB will review the accumulating data from the trial and assess whether there are any safety concerns or if there is enough evidence to demonstrate the effectiveness of the ALERT intervention. The interim analyses will also evaluate the potential benefits and harms of continuing or stopping the trial early.

The ALERT trial will conduct two interim analyses: one when approximately 50% of the total sample size is reached and another when approximately 75% of the total sample size is reached. The interim analyses will be conducted using pre-specified stopping rules and statistical criteria.

The use of interim analyses in the ALERT trial ensures that the trial is conducted in an ethical manner and that the safety and efficacy of the intervention are carefully monitored. If the intervention is found to be significantly effective, the trial may be stopped early to allow for earlier implementation of the intervention in clinical practice. Conversely, if the intervention is found to be harmful or ineffective, the trial may be stopped early to prevent further exposure of participants to the intervention.

### 5.5 Time-Points For Analysis

The ALERT trial will analyse outcomes at several time-points, including baseline, during intervention, post-intervention, and follow-up. Data will be collected at regular intervals to evaluate the impact of the intervention on primary and secondary outcomes such as perinatal mortality rates and maternal morbidity. The timing of data collection will vary depending on the specific outcome being evaluated. The aim of analysing outcomes at multiple time-points is to gain a comprehensive understanding of the short-term and long-term impact of the ALERT intervention on maternal and neonatal outcomes

### 5.6 Methods for Handling Missing Data

In the ALERT trial, missing data can occur due to various reasons, such as hospital withdrawal, or incomplete data collection. To address this issue, several methods will be employed to handle missing data, including:

1. **Complete case analysis:** In this method, only the data from participants with complete data at a given time-point will be included in the analysis. This method is simple but may lead to biased results if the missing data is not missing completely at random.
2. **Multiple imputation:** This method involves imputing missing data using statistical models based on the observed data. Multiple imputations are created, and the analysis is performed on each imputed dataset, and the results are combined. This method is more robust than complete case analysis and can handle missing data that is not missing completely at random.
3. **Last observation carried forward (LOCF):** In this method, missing data is imputed using the last observed value. This method is simple but may lead to biased results if the missing data is not missing at random.
4. **Sensitivity analysis:** Sensitivity analysis involves assessing the impact of missing data on the results by performing different analyses under different assumptions about the missing data. This method allows the researchers to evaluate the robustness of the study findings to missing data.

The choice of method for handling missing data will depend on the type and extent of missing data, the outcome of interest, and the assumptions about the missing data mechanism. By using multiple methods to handle missing data, the ALERT trial aims to ensure that the analysis is robust and that the results are not biased by missing data.

# STATISTICAL ANALYSIS PLAN



ALERT

## ACTION LEVERAGING EVIDENCE TO REDUCE PERINATAL MORTALITY AND MORBIDITY IN SUB-SAHARAN AFRICA (ALERT) TRIAL

The ~~golf~~ **First draft version**

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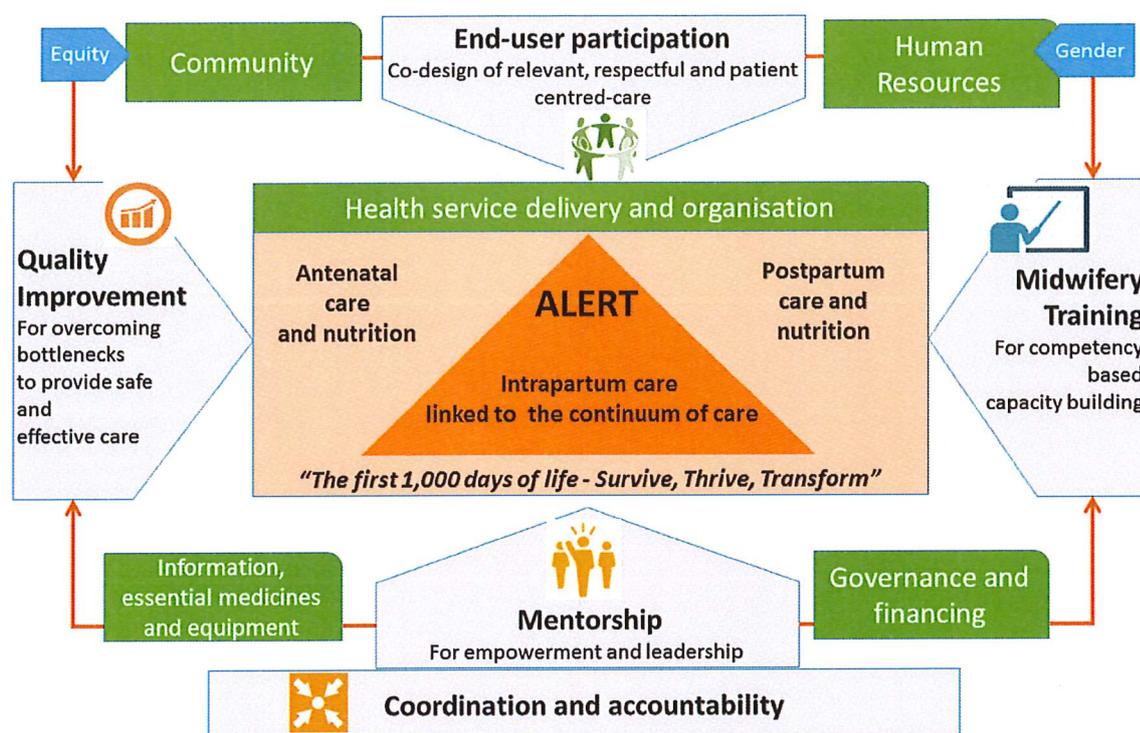
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# 1. STUDY BACKGROUND

The Action Leveraging Evidence to reduce perinatal mortality and morbidity in Sub-Saharan Africa (ALERT) trial was set to develop and evaluate a comprehensive and multilevel intervention covering intrapartum care and midwifery. The ALERT trial and interventions targeted hospital maternity units and included i) end-user participation of women, families, and midwifery providers to co-design the intervention; ii) in-service midwifery competency-based training; iii) empowerment and leadership mentoring of maternity unit leaders, and iv) quality improvement in the maternity ward, supported by district-based bi-annual coordination and accountability meetings (Fig. 1).

Figure 1: ALERT Conceptual Framework



## 2. STUDY OBJECTIVES

### 2.1 Specific objective

The ALERT trial had three specific study objectives:

1. To assess the ALERT intervention's impact in hospital maternity units on perinatal and maternal health outcomes, including women's experience of care.
2. To evaluate the process of implementation of the intervention to understand what works for whom and under what circumstances.
3. To conduct a cost-effectiveness analysis of the ALERT intervention.

### 2.2 Assessment of objectives

A summary of our approach to answering the objectives is given in Table 1. In this Analysis Plan, we give details for only objective one (details about other study objectives are given elsewhere).

**Table 1: Methods and data sources by study objective**

Objective	Methods	Data Source	Notes and links
<p>Obj 1. To assess the ALERT intervention's impact in hospital maternity units on perinatal and maternal health outcomes, including women's experience of care.</p>	<p>Trial profile                      Assessment of recruitment rate, engagement into the planned intervention e.g hospital engagement score and records completeness/incompleteness</p> <ul style="list-style-type: none"> <li>- Number of records entered the REDCap database in comparison to expected number of records (i.e. HMIS)</li> <li>- Number of incomplete records</li> </ul> <p>Descriptive statistics (counts, proportions, mean, median, SD, max, min)</p> <ul style="list-style-type: none"> <li>- Study outcomes split by intervention arm</li> <li>- Additionally, we will summarize the primary outcome split by both site and by intervention condition</li> </ul> <p>Regression models fitted for the primary, secondary and process outcomes.</p> <ul style="list-style-type: none"> <li>- Hospital-specific mixed-effects logistic regression models fitted separately for each hospital with random intercepts for randomization groups, followed by random-effects meta-analysis</li> <li>- Sensitivity analyses conducted by excluding outlier hospitals with data quality concerns</li> </ul> <p>Women's experiences of care</p> <ul style="list-style-type: none"> <li>- Descriptive statistics for each hospital, country (means and proportions)</li> <li>- Distribution of the of optimal care and number of interventions received.</li> <li>- Regression models to predict factors that determine optimal women's care</li> </ul>	<p>Perinatal e-registry                      Exit interviews</p>	<p>Our co-primary outcomes are Fresh stillbirth rate and in-facility early perinatal mortality (composite outcome).</p> <p>In-facility early perinatal mortality is defined as in-facility (fresh) stillbirth and 24-hour neonatal mortality</p>

Objective	Methods	Data Source	Notes and links
<p>Obj 2. To evaluate the process of implementation of the intervention to understand what works for whom and under what situation.</p>	<ul style="list-style-type: none"> <li>- Spatial analysis? – Health seeking behaviors vis-a-viz care seeking</li> </ul> <p>Trial profile</p> <ul style="list-style-type: none"> <li>- Table1: fidelity and implementation strength (from the REDCap fidelity data base)</li> <li>- Table2: Implementation in the hospital – to be developed and will include the improvement topics and change ideas by hospital.</li> <li>- Realist evaluation: <a href="#">Strengthening capacity in hospitals to reduce perinatal morbidity and mortality through a codesigned intervention package: protocol for a realist evaluation as part of a stepped-wedge trial of the Action Leveraging Evidence to Reduce perinatal morTality and morbidity (ALERT) in sub-Saharan Africa project - PubMed</a></li> </ul>	<p>WP7 – [WP7 published separate protocol]</p>	
<p>Obj 3. To conduct a cost-effectiveness analysis of the ALERT intervention.</p>	<p>Cost effectiveness analysis</p> <ul style="list-style-type: none"> <li>- Provider’s perspectives</li> <li>- Effects per outcomes</li> <li>- Incremental cost-effectiveness ratio (ICER)</li> <li>- Sensitivity analysis</li> </ul>	<p>Costing data</p>	

## 3. STUDY DESIGN

### 3.1 Overview

A stepped-wedge cluster-randomized trial (SWCRT) design with a nested process evaluation based on realist evaluation was employed in the ALERT trial. This design was selected as the interventions were administered at the hospital level, and a stepped-wedge design was chosen to mirror scale-up for policy buy-in and for statistical efficiency as larger cluster-level differences are anticipated.

### 3.2 Sample Size

The ALERT trial was carried out in 16 hospitals, with four hospitals selected in each of the four countries: Benin, Malawi, Tanzania, and Uganda. The selection criteria for the hospitals required them to have a minimum of 2500 births per year, which provided sufficient statistical power (75%-80%) to detect a 25% reduction in the in-facility early perinatal mortality rates with a 95% confidence interval. This also allowed for the assessment of other secondary outcomes such as maternal morbidity, hypoxic-ischemic insults, neonatal seizures, and more (see Figure 2).

Figure 2: Sample size and Power calculations for ALERT trial outcomes

	Baseline estimates	expected change	expected value	ICC	Power
<b>Perinatal e-registry (Cluster size 1500 births)</b>					
Stillbirth	1.0%	30%	0.7%	0.0024	83%
Stillbirth	1.0%	30%	0.7%	0.024	82%
Perinatal mortality	2.0%	25%	1.5%	0.0024	93%
Perinatal mortality	2.0%	25%	1.5%	0.024	92%
Perinatal mortality	1.4%	25%	1.1%	0.0024	70%
Hypoxic-ischaemic insults	4.0%	30%	2.8%	0.0024	100%
Hypoxic-ischaemic insults (sub-group)	4.0%	30%	2.8%	0.0024	79%
Neonatal seizures	0.5%	50%	0.3%	0.0024	78%
Cesarean section	15.0%	20%	18.0%	0.01	100%
Maternal morbidity	3.0%	30%	2.1%	0.5	99%
<b>Interviews post-partum (Cluster size 50 interviews)</b>					
Responsiveness	70.0%	15%	80.0%	0.1	97%
Mistreatment	4.0%	75%	1.0%	0.1	77%
Breastfeeding	60.0%	20%	71.0%	0.1	98%

**Source:** [Additional file 4 - Action leveraging evidence to reduce perinatal mortality and morbidity \(ALERT\): study protocol for a stepped-wedge cluster-randomised trial in Benin, Malawi, Tanzania and Uganda](#)

### 3.3 Randomization and blinding

The ALERT trial employed randomization, which was carried out by an independent statistician after obtaining consent from the participating hospitals. Stratification by country was used to enable random selection of four hospitals in each country (refer to randomization syntax in Figure 3a). Participants (hospitals) will not be blinded to the intervention, as with all training and quality improvement interventions. However, the women and their families receiving care at these hospitals will not be aware of the exact implementation stage of ALERT. The final sequence of ALERT intervention and data collection is illustrated in Figure 3b.

Figure 3a: Randomization syntax file

```
/*
This do-file is showing how to create a file with the randomized sequence
Date: 20200617
Author: Orsini N
*/

* Enter the country and hospital

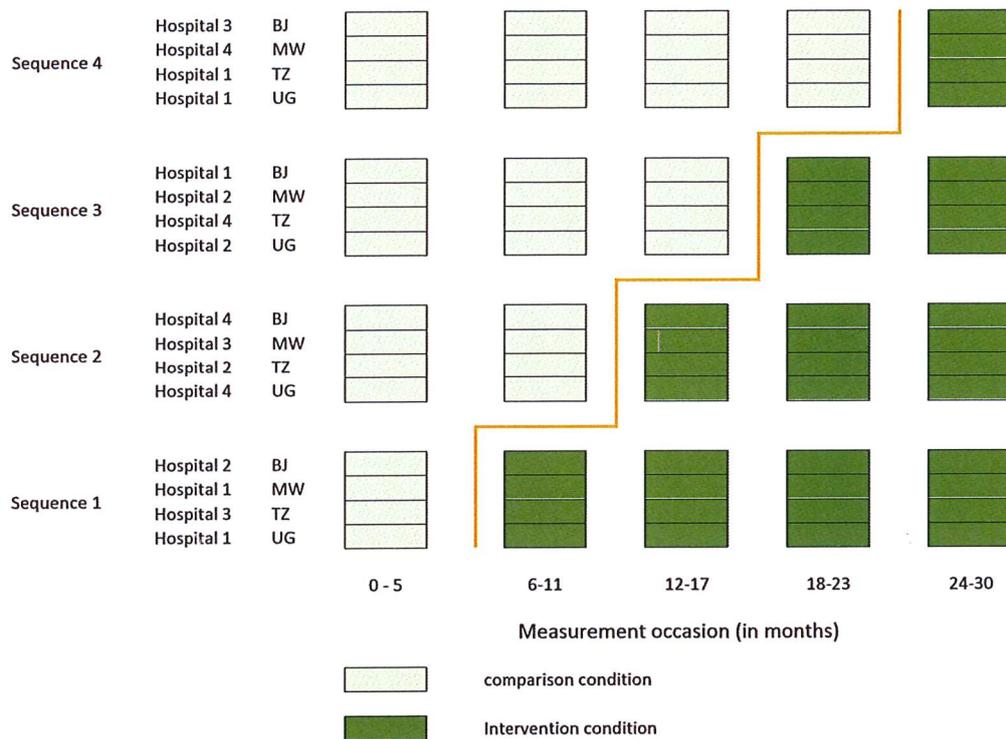
clear
input  str10 state  code str50  name
      Benin    1      "Centre National Hospitalier Hubert Maga"
      Benin    2      "Hopital de Zone Come-Grand Popo Houeyogbe Bopa"
      Benin    3      "Policlinique Bon Samaritan"
      Benin    4      "Centre Hospitalier et Dépt Zou-Collines"
      Malawi   1      "Mchinji Hospital"
      Malawi   2      "Mitundu Rural Hospital"
      Malawi   3      "Ntcheu D Hospital"
      Malawi   4      "St Gabriels Hospital"
      Tanzania 1      "Newala district Hospital"
      Tanzania 2      "Mkomaindo Hospital"
      Tanzania 3      "Ndanda Hospital"
      Tanzania 4      "Nachingwea Hospital"
      Uganda  1      "Bugiri General Hospital"
      Uganda  2      "Iganga Referral Hospital"
      Uganda  3      "Jinja Regional Hospital"
      Uganda  4      "Kamuli Mission Hospitla"
end

ssc install shuffle
gen rand = .

* Set the seed to reproduce the randomization
set seed 20200617

local list = "Benin Malawi Tanzania Uganda"
qui foreach s of local list {
    local j = 1
    shuffle 1 2 3 4, num noisily
    local seq = r(list)
    foreach r of local seq {
        replace rand = `r' if code == `j' & state == "`s'"
        local j = `j' + 1
    }
}
export excel using "results_rand_200617", firststrow(variables) replace
```

Figure 3b: Sequence of the ALERT trial intervention



### 3.4 Study assessments

To achieve objective one, we will conduct two assessments. Firstly, we will abstract trial data from hospital registries and enter it into a REDCap application. Secondly, we will conduct exit interviews with hospital clients to capture and evaluate their care-related experiences while in the hospitals.

## 4. STUDY POPULATIONS

### 4.1 Definition of populations for analysis

To ensure that the results of the ALERT trial are relevant to its intended audience, different populations will be used for analysis. The primary analysis will involve all women who gave birth in the 16 participating hospitals, regardless of whether they received the intervention or not. This will be used to assess the impact of the ALERT intervention on the primary and secondary outcomes and the process.

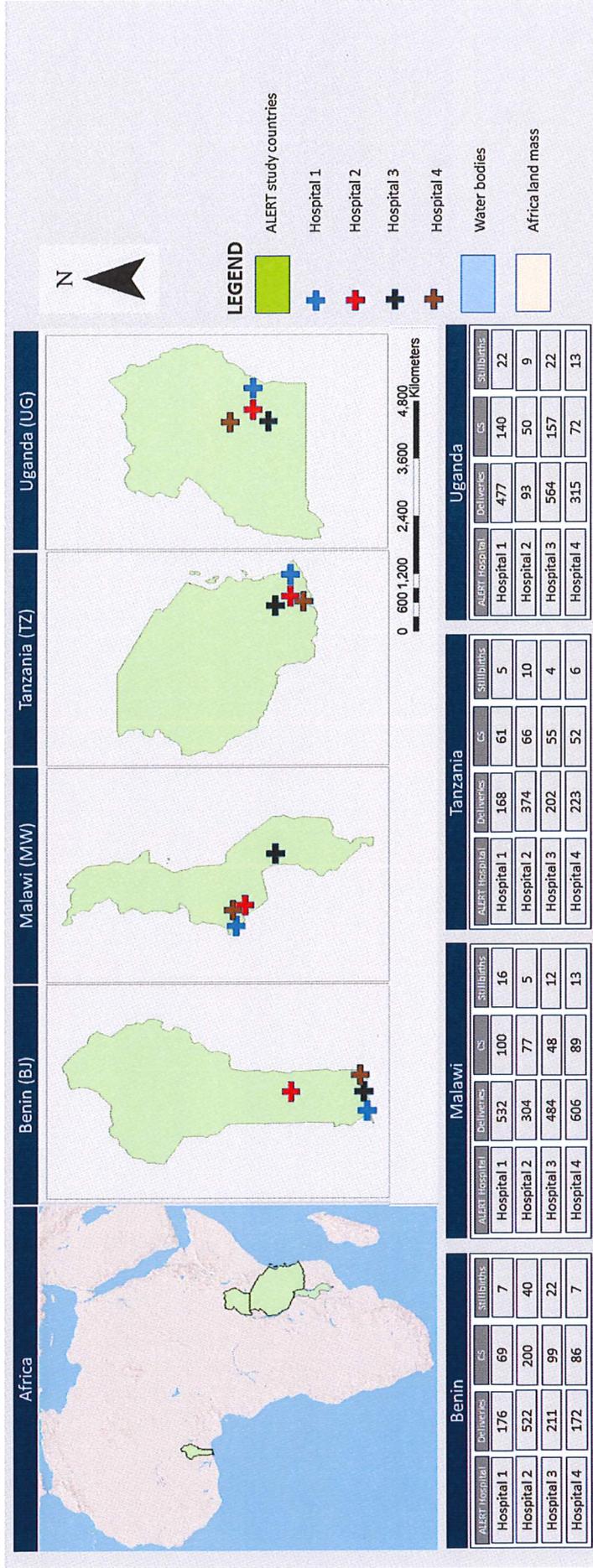
### 4.2 Intent-to-treat (ITT)

The intent-to-treat (ITT) analysis is a statistical method used in clinical trials that evaluates the effectiveness of an intervention in real-world conditions by analyzing all randomized participants, regardless of their adherence to the protocol or treatment received. In the ALERT trial, the ITT analysis will be used to assess the impact of the ALERT intervention compared to standard care. This method provides a more realistic estimate of the intervention's effectiveness in a real-world setting, but may underestimate the treatment effect as it includes all randomized participants, including those who did not receive the intended treatment, or received late or dropped out of the study.

### 4.3. Study sites and populations

The ALERT trial, which began its preparatory work in 2020, purposefully selected four hospitals in each country through consultations with the respective country's Ministry of Health (refer to Figure 4). All selected hospitals met the following criteria: (i) a minimum caseload of 2,500 births per year, based on trial sample size calculation; (ii) availability of cesarean section and blood transfusion services; (iii) preferably located in rural districts; and (iv) consisting of a mix of typical public but also private-not-for-profit (faith-based) hospitals. The inclusion of public and private-not-for-profit hospitals aimed to reflect the typical hospital landscape in sub-Saharan Africa and improve the generalizability of our results.

Figure 4: Map showing the ALERT trial countries and key baseline indicators for the study hospitals



## 4.4 Major protocol deviations

In any clinical trial, it is essential to adhere to the study protocol to ensure the validity and reliability of the results. However, sometimes deviations from the protocol may occur due to unforeseen circumstances or human error. In the ALERT trial, major protocol deviations will be defined as any deviation that affects the primary outcome or the safety of the participants.

Examples of major protocol deviations may include but are not limited to the following:

- Administration of the ALERT intervention to a non-intervention hospital
- Failure to administer the ALERT intervention in an intervention hospital
- Early termination of the intervention period
- Failure to obtain informed consent from participants
- Major errors in data collection or analysis

Any major protocol deviations will be reported to the trial steering committee, and appropriate action will be taken to address the deviation and mitigate its impact on the trial results. The impact of major protocol deviations on the primary outcome and safety of the participants will be assessed through sensitivity analyses.

## 5. STATISTICAL ANALYSIS

### 5.1. Study outcomes and indicators

The outcomes (primary, secondary and process) of the ALERT trial will be measured for all admitted mothers who deliver in the study hospitals and their pregnancy outcomes (livebirths, stillbirths, neonatal deaths) have been recorded in the e-registry from March 2021 to December 2023. The main study outcomes include;

#### Primary outcomes

1. **Fresh stillbirth rate:** Number of fresh stillbirths per 1,000 live and stillbirths
2. **Early perinatal mortality rate:** Number of fresh stillbirths and 24-hour neonatal deaths per 1,000 live and stillbirths (composite outcome)

#### Secondary outcomes

3. **Hypoxic-ischemic events:** Number of neonates with APGAR score less than 7 at 5 minutes per 1,000 live- and stillbirths
4. **Neonatal seizures:** Number of neonates diagnosed with seizures per 1,000 live- and stillbirths
5. **Cesarean section rate:** Number of cesarean sections per 100 live- and stillbirths
6. **Severe maternal morbidity:** Number of women with severe morbidities per 100 live- and stillbirths
7. **Responsiveness score:** Validated percentage score from exit interviews
8. **Respectful treatment score:** Validated percentage score measuring absence of mistreatment from exit interviews

#### Process indicators

9. **Fetal heart check on admission:** Percentage of babies with fetal check completed

10. **Intrapartum monitoring:** Fetal monitoring documented in first stage of labor
11. **Resuscitation for babies with APGAR < 7:** Percentage resuscitated
12. **Companion presence:** Companion present during labor or birth
13. **Breastfeeding within 1 hour:** Breastfeeding initiation within one hour.

## 5.2. Analytical approaches

The Makerere University School of Public Health and Karolinska Institutet, in collaboration with the London School of Hygiene & Tropical Medicine, will conduct statistical analysis of primary and secondary outcomes using Stata Standard Edition 18. The analysis will follow an "intention-to-treat" approach, comparing ALERT intervention clusters (hospital maternity wards) with comparison clusters where care is provided according to national standards. This approach will account for the complex clustering inherent in our design.

### *Data preparation*

The 30-month study period will be divided into five discrete six-month intervals. Time periods will be defined as:

#### **Stata**

```
egen time3 = cut(month), group(5) label
recode month (min/12= 0 "1-6 Months") (13/18=1 "7-12 Months") ///
(19/24=2 "13-18 Months") (25/30=3 "19-24 Months") ///
(31/36=4 "25-30 Months"), gen(time2) label(period)
```

Hospitals will be assigned anonymized identifiers using standardized Country-Hospital-Number format and sequential numerical coding from 1 to 16 for analytical purposes.

Treatment exposure will be constructed to reflect the stepped-wedge design where hospitals transition permanently from control to intervention status:

#### **Stata**

```
generate byte has_been_treated = 0
replace has_been_treated = 1 if time3 ==1 & group ==1 ///
| time3 ==2 & inlist(group, 1,2) | time3 ==3 & inlist(group,1,2,3) |
time3 ==4
```

Baby-level analyses will require data reshaping from wide to long format to accommodate multiple births per woman.

### *Descriptive analysis approach*

Baseline characteristics will be presented as frequencies and percentages for categorical variables, means with standard deviations for continuous variables with symmetric and bell-shaped distributions, and medians with interquartile ranges for very asymmetric distributions. Temporal trends will be visualized through monthly outcome rates by hospital to identify intervention effects and secular changes.

Sample Table 1: Distribution of Number maternal records collected from the e-registry

Clinics	Steps (2-month periods)											Total
	1	2	3	4	5	6	7	8	9	10	11	
1	1886	1280	1041	1249	1208	1336	1223	1181	945	1179	1340	13868
2	3653	2150	2250	2504	2846	2701	2604	2390	2204	2580	2404	28286
3	1360	769	830	664	799	930	872	803	765	731	768	9291
4	3138	2063	2055	1859	1683	2061	2227	2392	1888	1617	1841	22824
5	2775	1928	968	2258	2070	2015	2482	2366	2214	2034	2224	23334
6	2524	1624	1445	1602	1803	1772	1663	1595	1504	1569	1613	18714
7	1516	1293	1321	1310	1382	1476	1406	1343	1265	1357	1378	15047
8	3511	2307	1798	2179	2459	2693	2618	2593	2347	2220	2275	27000
9	2977	1816	1823	1529	1691	1728	1984	2006	1895	1535	1711	20695
10	3148	3634	2569	2730	3638	3652	3974	3983	3772	3941	4177	39218
<b>Total</b>	26488	18864	16100	17884	19579	20364	21053	20652	18799	18763	19731	218277

Control study period     Intervention study period

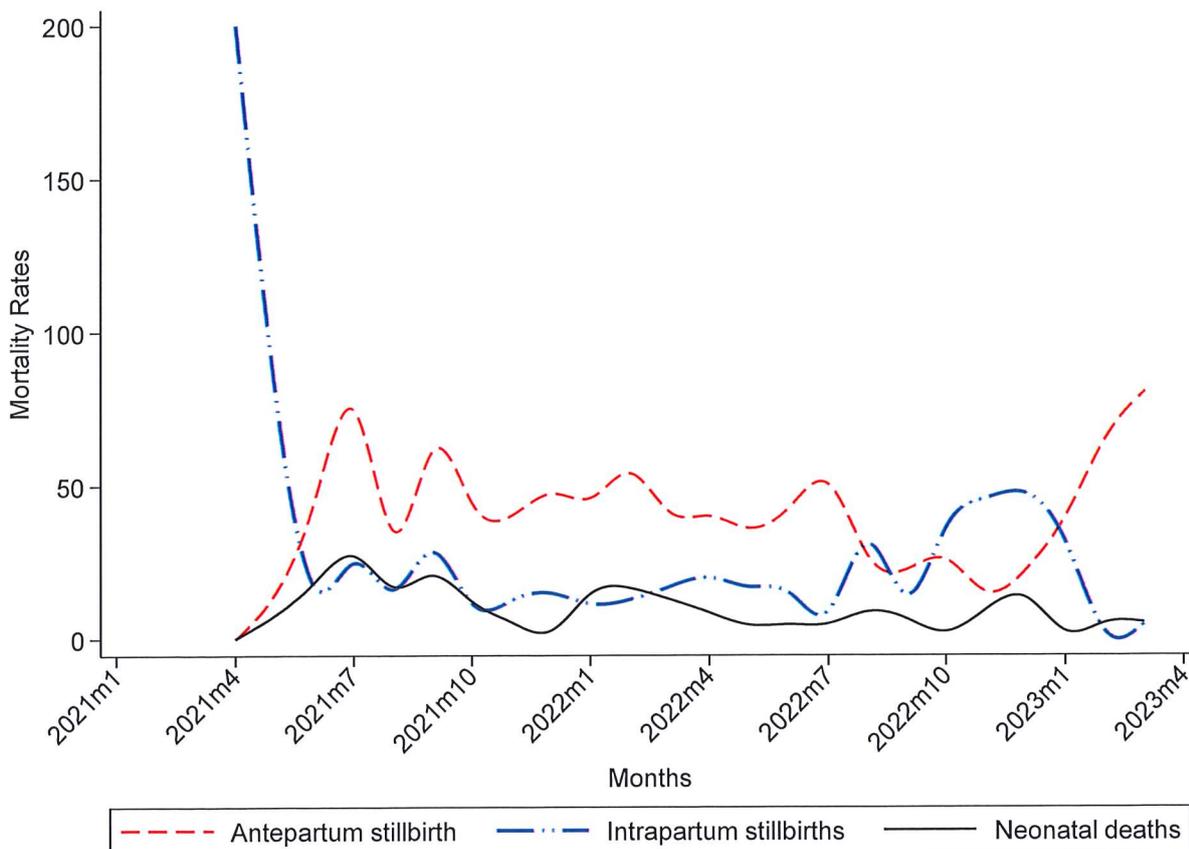
Table 2: Example of how number of admissions and data completeness could be presented

Hospital	Total number of records	No. completed records	Referral status and state at admission				Mode of delivery							Birth outcomes			
			Not referred	Yes – not in labor	Yes – in labor	Yes – after birth	SVD	C/S	Vacuum	Forceps	Breech	Singleton	Multiple births	Live birth	Fresh stillbirth	Macerated stillbirth	Neonatal deaths
Benin Hospital 1																	
Sequence 1																	
Sequence 2																	
Sequence 3																	
Sequence 4																	
Hospital 2																	
Sequence 1																	
Sequence 2																	
Sequence 3																	
Sequence 4																	
Etc.																	
Malawi Hospital 1																	
Sequence 1																	
Sequence 2																	
Sequence 3																	
Sequence 4																	
Hospital 2																	
Sequence 1																	
Sequence 2																	
Sequence 3																	
Sequence 4																	
Etc.																	

Table 2

Hospital	Variable	Control Arm	Treatment Arm
Benin			
Hospital 1	<ul style="list-style-type: none"> <li>• Number of records</li> <li>• Age</li> <li>• Status at admission</li> <li>• 1<sup>st</sup> Stage of labor mgmt</li> <li>• Mode of delivery</li> <li>• PNC checks</li> <li>• Initiation of breast feeding</li> <li>• Length of stay in hospital</li> <li>• N stillbirths</li> <li>• N livebirths</li> <li>• N neonatal deaths</li> <li>Etc.</li> </ul>		
Hospital 2			
Hospital 3			
Hospital 4			

Figure 5 – Example of graphical presentation of monthly mortality rates for Jinja regional referral hospital



### *Hospital-specific mixed-effects analysis*

The primary analytical approach will employ hospital-specific mixed-effects regression models to account for substantial heterogeneity across hospitals and healthcare contexts. We have chosen hospital-specific mixed-effects modelling over traditional pooled analyses to reflect the substantial heterogeneity between hospitals. We anticipate that hospitals operating in diverse healthcare contexts across four countries. While pooled approaches offer statistical efficiency, they risk masking important effect modification that could inform future implementation strategies.

Our modelling strategy recognizes that intervention effects may vary substantially across settings due to differences in baseline quality, staffing patterns, patient populations, and organizational cultures. By fitting separate models for each hospital, we preserve these nuances while enabling subsequent meta-analysis for overall effect estimation.

### **Box 1: Changes in methodological approaches**

#### **BOX 1: METHODOLOGICAL CHANGE FROM PREVIOUS VERSIONS (V1-V4)**

**Background:** Statistical analysis plans versions 1-4 (February 2021 - September 2023) specified Generalized Estimating Equations (GEE) as the primary analytical method for accounting for correlated data and clustering.

**Rationale for change to mixed-effects models:** During preliminary analyses conducted in early 2024, GEE models encountered substantial convergence failures and excessive computation times due to the complexity of the ALERT trial's data structure (16 hospitals, 4 countries, stepped-wedge design with varying implementation sequences).

**Decision process:** At the Data Safety and Monitoring Board meeting on May 28, 2024, the analytical team presented these computational challenges. Following discussion and review of methodological literature (particularly Hemming et al.'s guidance on cluster randomized stepped-wedge trials), the DSMB agreed to adopt mixed-effects regression models as the primary analytical approach.

**Advantages of mixed-effects models:** Enhanced computational tractability with reliable convergence, greater granularity in modeling hospital-level variation, appropriate handling of the nested data structure, and maintained interpretability of treatment effects while accounting for clustering and time trends.

### **Binary Outcome Specification**

For binary outcomes such as stillbirths, seizures, and cesarean sections, we will employ logistic mixed-effects models within each hospital. The mathematical specification follows:

#### **Model 1 – Logistic Mixed-Effects**

$$\text{logit}(P(Y_{ijk} = 1)) = \beta_0 + \beta_1 T_{ij} + \beta_2 \text{Time}_{ij} + \beta_3 \text{Group}_i + \alpha_i$$

Where:

- $Y_{ijk}$  = binary outcome for individual  $k$  in hospital  $i$  at time  $j$
- $T_{ij}$  = treatment indicator (1 = intervention, 0 = control)
- $Time_{ij}$  = time period effect
- $Group_i$  = randomization group fixed effect
- $\alpha_i$  = random intercept for randomization group

The Stata implementation will follow this syntax:

#### Stata

```
melogit outcome i.treat i.time i.group if hospn==j || group:, or
```

#### Continuous outcome specification

Patient experience scores from exit interviews require different modelling approaches given their continuous nature. We will employ linear mixed-effects models with the following specification:

#### ***Model 2 - Linear mixed-effects:***

$$Y_{ijk} = \beta_0 + \beta_1 T_{ij} + \beta_2 Time_{ij} + \beta_3 Group_i + \alpha_i + \varepsilon_{ijk}$$

Where:

- $Y_{ijk}$  = continuous outcome for individual  $k$  in hospital  $i$  at time  $j$
- $T_{ij}$  = treatment indicator (1 = intervention, 0 = control)
- $Time_{ij}$  = time period effect
- $Group_i$  = randomization group fixed effect
- $\alpha_i$  = random intercept for randomization group
- $\varepsilon_{ijk}$  = individual-level error term

Implementation:

#### Stata

```
mixed outcome i.treat i.round i.group if hospnam==j || group:, cov(ex)
```

#### ***Random effects considerations***

Our random effects structure acknowledges the stepped-wedge design's inherent clustering. Random intercepts for randomization groups account for the possibility that hospitals randomized to similar implementation sequences may share unmeasured characteristics influencing outcomes. This approach provides more appropriate inference than fixed-effects alternatives while maintaining computational tractability.

#### ***Systematic results management***

Managing results across 16 hospitals and multiple outcomes requires systematic data handling. We will employ Stata's postfile functionality to create comprehensive results databases. This approach

will ensure consistent results' capture while handling the convergence issues may that arise in complex mixed-effects models.

#### Stata

```
tempname memhold
postfile `memhold' hospn treat_beta treat_se pvalue using
"results.dta", replace

forval j=1/16 {
    capture noisily melogit outcome i.treat i.time i.group if hp==`j'
    || group:, or
    if _rc == 0 {
        matrix b = e(b)
        local beta = b[1,2]
        matrix v = e(V)
        local see = sqrt(v[2,2])
        local z = `beta' / `see'
        local pvalue = 2 * (1 - normal(abs(`z')))
        post `memhold' (`j') (`beta') (`see') (`pvalue')
    }
}

postclose `memhold'
```

#### Meta-analysis

Hospital-specific results will be combined using random-effects meta-analysis techniques employing the DerSimonian-Laird method:

$$\hat{\theta}_{\text{pooled}} = (\sum w_i \hat{\theta}_i) / (\sum w_i)$$

where  $w_i = 1/(\sigma^2_i + \tau^2)$ , incorporating within-hospital sampling variance ( $\sigma^2_i$ ) and between-hospital heterogeneity ( $\tau^2$ ).

Forest plots will display individual hospital effects alongside pooled estimates. I-squared statistics will quantify heterogeneity, with values above 50% indicating substantial variation requiring investigation.

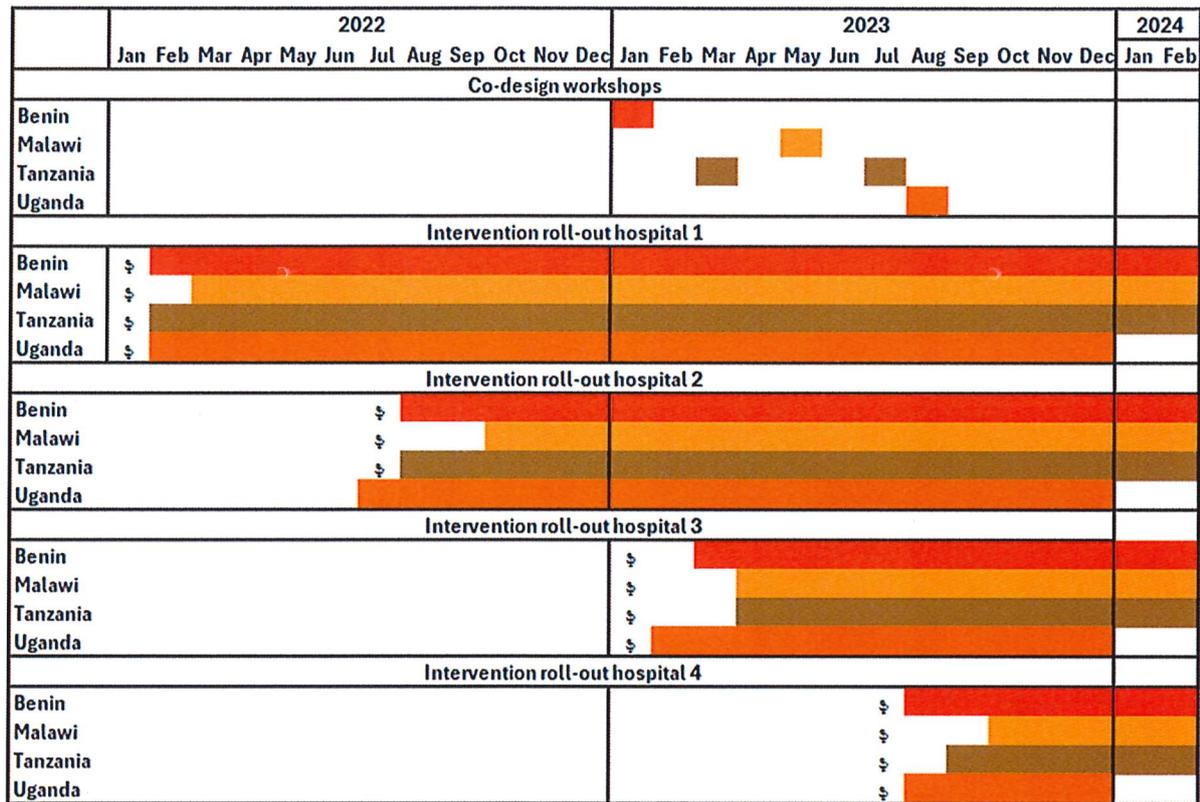
#### Sensitivity analyses

**Hospital exclusion:** Outlier Hospitals identified to have data quality concerns will be excluded to assess result robustness: After preliminary analysis, reviews with country teams will be held to contextualize unexpected patterns in hospital-specific results. Specific issues like governance crises and infrastructure deficits shall be probed and discussed. Decisions about exclusions for hospitals identified as outliers shall not solely be based on statistical outliers but also on contextual issues identified by country implementation teams.

**Stata**

```
melogit outcome i.treat i.time3 i.group if hospn==j & hospn!=23 ||
group:, or
```

**Alternative Time Specifications:** Multiple time period definitions will be tested including monthly and quarterly groupings.



⌘ planned start date

We have accommodated an alternative timeline in line with the actual implementation. All countries faced delays in engaging the hospitals, Uganda comparatively less. Our Per-Protocol analysis used the actual implementation. For per protocol analysis, a new variable reflecting the protocol implementation time shall be used instead of the “time” variable used for the intention to treat analysis (main analysis). Meaning the analysis syntax will essentially remain unchanged for the different outcome types.

**Missing Data:** Complete case analysis will serve as the primary approach. Multiple imputation will be employed where missingness exceeds 10%:

**Stata**

```
mi set wide
mi register imputed [varlist]
mi impute chained (logit) binary_vars (regress) continuous_vars =
predictors, add(10)
```

### 5.3 Statistical Inference

Two-sided hypothesis testing will be employed with  $\alpha = 0.05$ . Results will be interpreted considering patterns across related outcomes rather than individual p-values given multiple testing considerations.

Effect sizes will be presented as odds ratios with 95% confidence intervals for binary outcomes and mean differences with 95% confidence intervals for continuous outcomes. Forest plots will serve as primary results presentation, displaying individual hospital effects alongside pooled estimates and heterogeneity measures.

### 5.4 Missing data management

Electronic registry data will have minimal missing data for primary outcomes, warranting complete case analysis. Exit interview data may have higher missing rates requiring multiple imputation for key indicators with substantial missingness. Missing data patterns will be characterized systematically, with sensitivity analyses comparing complete case results with multiple imputation alternatives where missingness exceeds 10%.

### 5.5 Implementation

All analyses will employ Stata Standard Edition 18 with comprehensive documentation ensuring reproducibility. Results generation will use programmatic approaches with the table1 command for descriptive tables and custom programs for meta-analysis visualizations.

The analytical workflow will follow this structure:

#### **Stata**

```
// Data preparation
use allcountries16Apr24clean.dta, clear
[data management procedures]

// Hospital-specific analysis
forval j=1/16 {
    [individual hospital analysis]
}

// Meta-analysis
metan treat_beta treat_se, eform random [options]
```

## Annex: DSMB composition and tasks

We recruited the following members with a good distribution of experience (1 Obs&Gyn with trial experience, 1 statistician, 1 perinatal epidemiologists)

Zahida Qureshi [zahida@qureshi.co.ke](mailto:zahida@qureshi.co.ke); <https://profiles.uonbi.ac.ke/qureshi/>

Erik Lampa [erik.lampa@medsci.uu.se](mailto:erik.lampa@medsci.uu.se) ; <https://katalog.uu.se/empinfo/?id=N11-1203>

Jennifer Hall [jennifer.hall@ucl.ac.uk](mailto:jennifer.hall@ucl.ac.uk); <https://iris.ucl.ac.uk/research/personal?upi=JAHAL68>

### Rational for DSMB

ALERT is a stepped-wedge cluster (hospital) randomised trial to evaluate a public health intervention composed of four components of end-user participation, midwifery training, quality improvement and mentoring for the head of maternity (find the summary attached). It is funded by Horizon 2020 and will be implemented in Benin, Malawi, Tanzania and Uganda. The intervention aims to reduce perinatal mortality and improve intrapartum care including user responsiveness.

ALERT is not a clinical trial. ALERT will implement and evaluate the effect of non-clinical interventions such as training and quality improvement. However, also non-clinical intervention through changes in practises may have clinical consequences. For example, changes in clinical practice may increase in Caesarean section rates, those clinically justified but potentially also some without clear indication. This is an unintended consequence and as Caesarean section bear risks for mother and their babies, Caesarean section rate will need to be monitored. Moreover, the main study outcomes include stillbirth and early neonatal mortality rate. Thus, the establishment of a DSMB is indicated.

### The DSMB should have the following tasks

- To review the trial procedure including randomization and attainment of outcomes (April 2021)
- To review protocol compliance and enrolment at month 12 of the trial / June 2022 (Interim review of protocol compliance)
- To review the effects of the intervention at month 12 and months 24 of the trial, including a check on futility, safety and interim impact to advise on continuation (June 2022 & June 2023,)
- To advise the PI (Claudia Hanson) and the ALERT Trial Management Team

The ALERT trial management team will include the statistician Nicola Orsini (Karolinska Institutet) Joseph Akuze (Makerere University), Peter Waiswa (PI Uganda, Makerere University), Jean-Paul Dossou (PI Benin, CERHHUD), Effie Chipata (PI Malawi, College of Medicine) and Andrea Pembe (PI Tanzania, MUHAS).

April 2021: The ALERT trial management team will prepare a full trial protocol. We will invite for a meeting either in-person or using teleconferencing to discuss the protocol and establish the further working modalities. During this meeting stopping rules and procedures for severe adverse events will also be discussed as well as the forms for sharing unexpected findings and events.

June 2022: We will prepare a report on protocol compliance and share data together with an analysis do-file to enable the DSMB to review the enrolment and safety.

June 2023: We will prepare an updated report on protocol compliance and share data together with an analysis do-file to enable the DSMB to review the enrolment and safety. The interim and the final analysis will be blinded.

We expect the team to be available for the DSMB at the indicated timepoints and expect at each timepoint about 8 hours to be spend on this assignment including reading and commenting on the report and joining the virtual or actual meetings. Potential travel expenses will be taken care of by the project. We have not foreseen any remuneration.