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PEDIATRICS

Post-Discharge Mortality Prediction in Sub-Saharan Africa

Post-Discharge Mortality Prediction in Sub-Saharan Africa

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POTENTIAL CONFLICTS OF INTEREST: The authors declare no conflicts of interest.

ABBREVIATIONS.

TABLE OF CONTENTS SUMMARY: This study shows the burden of postdischarge mortality in Southern Mozambique, identifying predictors which could efficiently stratify children with higher risk of dying following a hospital discharge.

WHAT'S KNOWN ON THIS SUBJECT: Post-discharge mortality is an important contributor to child mortality, ranging between 3.3% and 13%, although it is poorly recognized. No predictive models of post-discharge mortality among all causeadmissions in resource-constrained hospitals or among infants have been developed to date.

WHAT THIS STUDY ADDS: Predictive models presented in this study could be applied at hospital discharge and children at risk of dying could be identified through their use. This could allow designing a better post-discharge planning, health education to the families and follow-up care.

AUTHORS´CONTRIBUTIONS

Dr. Madrid conceptualized and designed the study, cleaned and analyzed data, interpreted data set results, drafted the initial manuscript, and reviewed and revised the manuscript. Drs. Bassat and Cousins conceptualized and designed the study, interpreted data set results, and critically reviewed the manuscript for its content.

Drs. Alonso, Menéndez, Macete, Sacoor, Varo, Sitoe, Acacio, Nhampossa and Sigaúque, coordinated and supervised data collection, and critically reviewed the manuscript for its scientific content.

Sergio Massora and Dr. Inacio Mandomando were responsible for laboratory procedures, quality of CISM laboratories facilities and interpretation of results. Mr. Quintó and Miss Casellas led the data analysis, interpreted data set results, and critically reviewed the manuscript.

All authors approved the initial manuscript (and subsequent versions) as submitted and agreed to be accountable for all aspects of the work.

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ABSTRACT

Background: Although the burden of post-discharge mortality (PDM) in low-income settings appears significant, no clear recommendations have been proposed in relation to follow-up care after hospitalization. We aimed to determine the burden of paediatric PDM and develop predictive models to identify children at risk of dying following discharge.

Methods: Deaths after hospital discharge among children aged <15 years in the last 17 years were reviewed in an area under demographic and morbidity surveillance in Southern Mozambique. We determined PDM over time (up to 90 days) and derived predictive models of PDM using easily collected variables upon admission.

Results: Overall PDM was high (3.6%), with half of deaths occurring in the first 30 days. One primary predictive model for all ages included young age, moderate/severe malnutrition, history of diarrhoea, clinical pneumonia symptoms, prostration, bacteraemia, positive HIV status, rainy season and transfer or absconding, with an area under the curve (AUC) of 0.79 (0.75-0.82) at day 90 after discharge. Alternative models for all ages including simplified clinical predictors had a similar performance. A model specific to infants <3 months identified as predictors: being a neonate, low WAZ score, breathing difficulties, hypothermia or fever, oral candidiasis and history of absconding or transfer to another hospital, with an AUC of 0.76 (0.72-0.91) at day 90 of follow-up.

Conclusions: Death following discharge is an important although poorly recognized contributor to child mortality. A simple predictive algorithm based on easily recognizable variables could readily identify most infants and children at high risk of dying after discharge.

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BACKGROUND

The last 25 years have witnessed a significant (49%) reduction in under-5 child mortality globally¹, but an insufficient one to meet the two thirds reduction objective set by the fourth millennium development goal, particularly among many low-income countries $(LIC)^2$.

In the last decades, algorithms for diagnosis and treatment of sick children have been implemented in order to address the management of disease during the acute phase and have contributed to improve child survival³. Such guidelines and recommendations, however, have historically failed to address the days immediately following hospitalization, a critical period for child survival⁴.

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l children, albeit scarce, hav Contrary to what occurs in industrialized countries, where post-discharge mortality (PDM) is limited to certain small, high-risk groups^{5, 6}, children in LIC appear to be at increased risk of mortality following hospitalisation for any illness^{4, 7-13}. Previous studies among admitted children, albeit scarce, have estimated risk of PDM to range between 3.3% and $13\%^{9, 14, 15}$. For specific diseases, PDM risks have ranged between 2.0-2.6% for malaria^{16, 17}, 2.9 to 7% for diarrhoea^{18, 19}, 2.7-11.6% for anaemia²⁰, 1-15% for pneumonia^{15, 21}, and 2.8% for invasive bacterial infections⁷. Most of these deaths have been described to cluster in the first 30 days^{13} and main predictors of PDM include a history of previous hospitalizations, young age, HIV infection and hospitalizations related to malnutrition or pneumonia⁴.

A rigorous follow-up of all hospital discharged children would be unfeasible and unaffordable for resource-constrained settings. Thus, early identification of vulnerable children appears essential to design more targeted interventions to prevent $PDM¹³$. Improving the discharge process and post-discharge care will be a critical step to further

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continue reducing child mortality⁴. We therefore aimed to determine the burden of paediatric PDM in a semi-rural area from southern Mozambique, to identify predictors of mortality following discharge and to derive models that could efficiently stratify children according to PDM risk.

METHODS

Study site and population

population (45% <15 years of age). In
5/1000 live births²². Manhiça Health I
weillance system (DSS) in the area, ar
1SS) has been implemented for nea
District Hospital (MDH) and five add
ring standardized morbidity data This study was conducted in Manhiça, southern Mozambique, a semi-rural setting with a predominantly young population (45% <15 years of age). In 2015, the national under-5 mortality rate was $78.5/1000$ live births²². Manhica Health Research Centre (CISM), runs a demographic surveillance system (DSS) in the area, and a paediatric morbidity surveillance system (MSS) has been implemented for nearly two decades at the neighbouring Manhiça District Hospital (MDH) and five additional peripheral health posts, accurately capturing standardized morbidity data for \sim 3000 admissions and >75,000 annual outpatient visits, annually. Human immunodeficiency virus (HIV) prevalence in the area is among the highest in the world²³, with adult community prevalence peaking at $40\%^{24}$. Vertical transmission of HIV has been estimated at around $9\%^{25}$ and contributing 10% to the under-5 mortality nationally²⁶. A detailed description of CISM and study area can be found elsewhere^{27, 28}.

Study design and definitions

A retrospective cohort study of children <15 years discharged from MDH for a 17 yearlong period was conducted, using the DSS and MSS databases. We analysed burden of PDM over three different time-periods: First (1-30 days), second (31-60 days) and third month (61-90 days) post-discharge. PDM among infants <3 months old was analysed

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Rev separately to check whether identified predictors differed from those of the older cohort. Only children living in the study area were included. We used a single-discharge approach, considering the first admission as reference admission, and not considering ulterior re-admissions within the following 90 days to avoid re-starting the period at risk every time. Post-discharge death was defined as a death occurring >24 hours after and within 90 days of discharge from MDH; community death as a death ocurring outside a health facility within 90 days after discharge; and facility death as a death ocurring at any health facility within 90 days following discharge from MDH. Chidren dying during a readmission within the follow-up period were considered as a hospital death within follow-up (supplementary figure S1). Supplementary table S1 summarizes other relevant definitions.

Morbidity surveillance system.

Morbidity surveillance data routinely collected for all children <15 years during the study period were analysed, including clinical data, basic laboratory investigations (malaria microscopy, haematocrit and glycaemia) and International Statistical Classification of Diseases-based diagnoses (ICD-10) since 2003. All diagnoses prior to July 2003 were coded according to a list of codes created by the CISM since 1996. Outcome and medications prescribed were also analysed. For admitted children, blood culture results, systematically performed for all children \leq years and, in older children, in those with severe disease, were available. HIV status information, although not routinely collected, was available for those patients with suspected immunosuppression.

Demographic Surveillance System

CISM's DSS, which started in 1996, now covers the entire Manhiça District (2380 km^2) ; total population of ~183000 inhabitants). DSS captures socio-demographic data and other major events like migration, marital status, pregnancy and outcome results, births and deaths and is updated twice annually. During periods in which the entire district was not covered, analysis only included inpatients being part of the DSS. Through a unique identifier number, DSS and MSS databases can be linked.

Data management and data analysis

is performed to model events within 90
as used as date of entry to the study wh
d of follow-up period (90 days after d
riable with a high proportion of mis
confounder, a "missing or unknown"
study objectives, data analysis A survival analysis was performed to model events within 90 days of discharge. The discharge date+1 day was used as date of entry to the study whilst post-discharge death, loss to follow-up or end of follow-up period (90 days after discharge), considered the exit time. For each variable with a high proportion of missing values (<15%) but suspected to be a strong confounder, a "missing or unknown" category was created. In order to achieve the study objectives, data analysis was split in two parts: a) *Determining burden and identifying associations*: Descriptive statistics were calculated for all explanatory variables. Kaplan-Meier (KM) curves were produced for all categorical predictors to look for differences in survival with different values of the predictor. Associations between potential predictors and risk of death after discharge were explored in univariable Cox regression models. *b) Selecting and validating predictive models:* The dataset was randomly split into two subsets (training set containing 80% of data and the validation set with the remaining 20%) which were then compared to confirm that there were no important differences between the two subsets. Those predictors showing evidence of an association (p-value ≤ 0.05) with the outcome in a univariable analysis were selected for potential inclusion in a multivariable Cox regression model (primary model). Three additional models were also examined based on their suitability for different contexts: Model 2), which uses the primary model as reference, but includes only variables with minimal costs; Model 3) based exclusively on clinical variables collected on admission; and Model 4) predictors of PDM restricted to infants <3months. The area under the curve (AUC) was plotted for each model over time using the training set (formula: $H(t) = H_0(t)$ x $exp(b_1X_1 + b_2X_2 + b_3X_3 + ... + b_nX_n)$ b_kX_k). Confidence intervals of AUC were estimated by 1000 bootstrap replicates using the bias-corrected percentile method²⁹. As there could be more than one admission for some children, the models described above were estimated taking into account withinchild clustering. Analyses used Stata Statistical software (*Release 15*). Graphical representation of AUC curves was done in R (R Core Team; 2017) using the *survivalROC* package.

Ethical considerations

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C This study examined data collected in the context of routine clinical practice. DSS and MSS ongoing in the study area have been approved by the National Ethics Committee of Mozambique.

RESULTS

Overall characteristics of study population

From $1st$ January 2000 to $31st$ December 2016, 58990 inpatient records were checked of which 29574 (50%) were initially excluded (figure 1). 3097 observations of children living in study area readmitted within follow-up (supplementary figure S1), hospital deaths (2.5%, 662/26319 of remaining children) and 25 deaths in the first 24 hours $(0.1%) were also excluded (figure 1). Thus, 25632 inpatient records of 18023 children$ <15 years old admitted to MDH were included in the analysis (Table 1). 2055

observations of 2049 infants <3months were also analysed separately (supplementary table S2).

Incidence of post-discharge deaths and potential predictors of PDM

During the 90-day follow-up period, 935 (3.6%) deaths after discharge occurred among the 25632 admissions, with 783/935 (83.7%) occurring at the community level and 488/935 (52.2%) within the first 30 days of discharge. Median time to death was 28 days (IQR 11-53). The risk of post-discharge deaths varied over time (figure 2A and supplementary figure S2) and by age (figure 2B) similarly to the risk of inpatient mortality (figure 2C).

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validation
3 presents the comparison of the train
 Forty-five variables were tested for their univariable association with PDM (table 1). Infants <3months were more likely to die than older children (figure 3A; supplementary table S2). Similarly, poorer nutritional levels were clearly linked to PDM (figure 3B).

Predictive models and validation

Supplementary table S3 presents the comparison of the training and validation sets. Nineteen variables were associated with higher risk of PDM in the multivariate analysis (primary model, table 2). Infants less than 3 months had the highest rate of PDM and this decreased with increasing age $(p<0.001)$ (supplementary table S4). The rate of PDM varied between rainy and dry seasons (HR 1.22, 95% CI 1.03-1.43). Severe acute malnutrition (SAM) (HR 3.26, 95% CI (2.08-5.12) as well as other 13 clinical variables were associated with PDM. Children with a positive blood culture (HR 1.68, 95% CI 1.33-2.12) and a positive HIV test (HR 1.77, 95% CI 1.07-2.91) also had a higher rate of death during the follow-up period. Absconding (HR 5.23, 95% CI 4.22-6.50) or referral to a higher level of health care (HR 4.48, 95% CI 3.31-6.05) were clearly

associated with death. Children with a malaria diagnosis had a lower risk of PDM (HR 0.44, 95% CI 0.36-0.54). The AUC for this model was around 0.80 during the 90 days follow-up period, being, at day 90, 0.79 (95% CI 0.75 - 0.82). At a HR cut-off point of 1.08, it had a sensitivity of 80%, specificity of 60% and a positive predictive value (PPV) of 6.9% (figure 4 and table 3). According to this model, 1.9% of all discharges will die in the first 90 days after discharge.

3A. *Model 2*, which excluded blood
formed similarly to the primary mode
ter discharge. *Model 3*, which include
e, with an AUC ~0.75 during the whole
ncluded variables such as breastfeed
ional status, on account of the e Time-varying AUC of the two additional simplified models (table 2), is compared in supplementary figure S3A. *Model 2*, which excluded blood culture but maintained minimal cost tests, performed similarly to the primary model, with an AUC around 0.80% until 60 days after discharge. *Model 3*, which included only clinical variables, performed slightly worse, with an AUC ~0.75 during the whole period. *Model 4*, limited to infants <3months, included variables such as breastfeeding and weight-for-age (WAZ) to assess nutritional status, on account of the excess of missing height data. Neonates appeared to have the highest risk of PDM among this age group (supplementary table S4). This model had an AUC at day 30 of 0.84 (95% CI 0.72 - 0.91) and at day 90 of 0.76 (95% CI 0.72 - 0.91). At a HR cut-off point of 0.5, it had a sensitivity of 79%, specificity of 53% and a PPV of 12.8% (table 3 and supplementary figure S3B). AUC and model's characteristics at probability cut-offs ensuring sensitivity of >80% are shown in table 5. CI of AUC between training and validation set and between primary model and the other models overlapped, meaning no significant differences between them were found.

DISCUSSION

This analysis, based on more than 20000 hospital discharges and 935 post-discharge deaths, is the largest study to date evaluating PDM in the first three months following hospital discharge from a rural district hospital in a LIC, and represents a systematic approach to ascertain predictors of PDM in a resource-constrained environment.

The cumulative three-month post discharge mortality found in this study (3.6%) is lower than that reported in the 1990s from other African settings, where incidence risk was estimated between 6.1% and $13\%^{14, 15}$, but similar to other more recent PDM studies^{7, 9, 10, 18}. Importantly, inpatient mortality found in our cohort (2.5%) aligned closely with that reported in other settings $^{13, 14}$.

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the most critical period for survival¹⁰

for all ages changed over time. PDM

subsequently declining. This trend cantality, as this was progressiv These findings highlight that risk of dying is greatest in the first 30 days immediately following discharge^{7, 10}, the most critical period for survival^{10, 15, 18}. Overall, trends of post-discharge mortality for all ages changed over time. PDM rates increased over the period from 2000-2010, subsequently declining. This trend cannot be explained by an increase in inpatient mortality, as this was progressively decreasing since 2001 (figure 2C). One could speculate that variations in the epidemiology of a single disease may have played a significant role in PDM variations (Supplementary figure S4). For instance, malaria used to be highly endemic in the area at the beginning of the study period, with a subsequent declining trend observed from the year 2005 onwards, trends inversely proportional to those of the PDM curve. Children admitted with malaria, readily treatable and with a rapid recovery, are probably less likely to be associated with post-discharge complications. This could partly explain why a diagnosis of clinical malaria has been shown to be an overall protective factor against PDM (HR 0.44, 95% CI 0.36-0.54) a finding also reported in Kenya⁹. Similarly, although HIV incidence is now much higher in the area than it was a decade ago, the increasing uptake of

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sk of dying in the first antiretroviral drugs probably implies a better control of this infection on PDM in recent years when more admissions due to HIV have been registered. Trends in the proportion of children admitted with other diseases with greater incidence of post-discharge deaths, such as SAM, showed an overall tendency parallel to the PDM curve over time. However, the declining trend of PDM seen at the end of the study period cannot be exclusively explained by malaria, HIV or SAM trends. The progressive introduction of other life-saving interventions, such as vaccination against Hib (2009), pneumococcus (2013), and rotavirus (2015) have reduced hospital admissions and, may also have contributed to these decreasing trends. PDM trends over time among infants <3 months differed slightly from other age groups, with a peak in 2003, and stable rates thereafter. In this highly vulnerable age group, effective interventions specifically designed to reduce mortality, have not been fully implemented in the study area.

The primary model, including all available useful variables performed remarkably well, particularly to predict risk of dying in the first month following discharge. This model predicts that 1.9% of children discharging from the hospital will die in the first 90 days following discharge. Applying a score based on this model to a population similar to ours and using a HR cut-off of 1.08, roughly 80% of children likely to die following discharge would be identified and the referral population would have a mortality risk of approximately 7%. This model however includes blood culture results, an expensive determination that requires laboratory infrastructures seldom available in poor settings. The two alternative simplified models, including more easy-to-collect lab results (Model 2) or only clinical variables (Model 3) seem more applicable, and still showed a good predictive capacity for PDM. Importantly, model 4, developed for infants <3 months old, uses only few variables, all of them easily and readily obtainable in most

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resource constrained contexts and could identify 0.85 of infants <3months at risk of dying in the first days after discharge and nearly 0.80 during the remaining follow-up period.

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er number of admissions and Severe acute malnutrition is a recognized predictor of $PDM^{8-10, 13, 18}$ as our models confirm. The chronicity of malnutrition, the fact that it can predispose to an array of coinfections and complications, and its association with HIV infection in Manhica³⁰, may all contribute to explain its associated prolonged mortality risk. On the other hand, children with unknown nutritional status had higher risk of dying compared to wellnourished children. This may be partly explained in cases of severe disease, where the severity of the child upon admission did not allow collecting anthropometric measurements. HIV positivity, a clear predictor of $PDM^{13, 20}$, remained in our model a strong independent predictor, even after adjusting for malnutrition. Rainy season was also associated to higher risk of PDM as more children are admitted during this season, likely due to the greater number of admissions and severe disease occurring in this season.

In this series, type of outcome at discharge, and particularly being transferred or absconding from hospital, were the greatest predictors of PDM. In this setting, children are usually transferred to a higher-level health facility whenever they are very sick or when they require a more specialized evaluation or supportive care, justifying their greatest PDM risk. Children absconding against medical recommendations (3.1% of the study sample) had an extremely high risk of post-discharge mortality, and represent one out of every five deaths in our series. In Manhiça, absconding is a cultural and financial phenomenon, typically occurring when families anticipate a bad outcome and prefer their children to die at home, additionally sparing costs associated with the transport of a corpse. Socio-behavioural studies addressing this phenomenon and the perceptions of health professionals of its serious consequences are needed.

The majority (83.7%) of deaths following hospital discharge occurred at the community, in the absence of any further contact with the health system. A study investigating cause of death in Manhiça using verbal autopsies documented that 53% of all paediatric deaths occurred at home²⁸. These alarming figures reflect the generalized challenges in access to care, which become even more blatant following hospital discharge.

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ly that risk can be red Effective models for identifying children at risk of PDM should take into consideration the existing resources but consider illness as a continuum transcending the information that the admission snapshot can provide. A score or similar algorithm to that proposed by IMCI, based in predictive models and applied at discharge, could pinpoint children requiring a more rigorous follow-up after hospitalization. However, the identification of high risk does not imply that risk can be reduced. Future research should consider validation of these models in different contexts and prospectively assessing their accuracy to identify children at risk of dying after discharge in resource-constrained settings. Once identified, these children at higher risk of PDM could benefit from strategies to prevent post-discharge death and these strategies should especially focus in the first 30 days after discharge as it is the period with the highest risk of PDM. Community-based interventions driven by community-health workers consisting in pre and post-natal home visits, supporting low birth-weight (LBW) infants and sepsis case management, facilitating referral in case of need have reduced neonatal and infant mortality in several countries³¹. Although these interventions have not been explored in children after a hospital admission, their impact reducing PDM could be similar.

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Alternative strategies, utilizing the prophylactic use of antimicrobials in those children at high risk should also be explored. A recent clinical trial conducted in Kenya exploring the efficacy of daily post-discharge co-trimoxazole prophylaxis in children admitted with complicated SAM without HIV found however no reduction in mortality during the first year after admission 32 .

 Continued investment in child mortality data collection and understanding circumstances of paediatric death following a hospital discharge is needed in order to design innovative, effective and feasible strategies to reduce the risk of childhood preventable deaths after hospitalization.

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Initiations, including its retrospective n

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Interact that all clinical pr This study has several limitations, including its retrospective nature. Selection bias may arise due to the fact that almost half of children admitted to MDH between 2000 and 2016 were excluded and this might affect the representativeness of the study sample. Another limitation includes the fact that all clinical predictors were collected at the time of admission, and some these clinical variables may have changed throughout the hospitalization. We decided to use a single-discharge approach within 90 days of follow-up to avoid double counting time at risk, excluding observations of children readmitted during the follow-up period. This strategy may have resulted in a clearer picture of the true community PDM but also in an underestimation of the likely higher real-life true incidence. Another factor potentially underestimating PDM is the exclusion of deaths occurring in the first 24 hours after discharge, as they were considered as hospital deaths. However, they merely represented <0.1%. The inclusion of children who were transferred or absconded from the hospital may be overestimating the incidence of PDM since they were not officially discharged, but this is an extremely frequent occurrence in African settings, and needs to be taken into consideration, particularly in the light of the strength and magnitude of the statistical associations found. Importantly, we could not assess the role of LBW in infants as a likely risk factor for PDM, since this information was not available. On the other hand, the low specificity and positive predictive value found could compromise feasibility of interventions to prevent PDM, as a high number of children would be classed as high risk of dying after discharge. However, these models would allow the identification of the majority of children with risk of PDM. Finally, our predictive models lack an external validation.

Conclusion

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en at the household level This study highlights the importance and oversight of post discharge mortality as a significant portion of the childhood mortality pie. Simple models including predictors easily collected with minimal cost, such as those presented in this article, need to be prospectively validated in different circumstances and settings. Specific interventions targeting children identified to be at higher risk and guaranteeing their adequate followup at the hospital or even at the household level could possibly increase their survival possibilities. Implementation of such strategies could prevent avoidable deaths, especially among neonates and infants who suffer the highest burden of PDM.

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Page 2¹ **Figure 1. Study profile**

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Figure 2: Mortality trends over time during the 17 year-long study period

- A) Yearly incidence of post-discharge mortality for all ages.
- B) Yearly incidence of post-discharge mortality by age group.
- C) Yearly incidence of inpatient mortality for all ages

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The time–varying^{Confidential - Not for Circulation ve of primary model^{e 24 of 46}
The time–varying area under the curve of primary model} **comparing training and validation sets**

Table 1. Socio-demographic, clinical characteristics and univariate analysis of predictors on admission associated to post-discharge mortality in southern Mozambique, based on 25632 observations and 935 post-discharge deaths.

123456789

ᶧSD: standard deviation. ᶤWHZ: Weigh-for-height. See definitions in Table S1. ᵝIt refers both to community deaths and deaths in a readmission during follow-up period. ^hPercentage represents risk among children with same characteristics. Ψ HR: Hazard ratios. HR and confidence intervals were derived from a Cox regression model.*Confidence intervals. ⁰ P-value was derived from Wald test. BCS: Blantyre coma score. ^{*}History of current disease reported by the child carer. ⁹Hospital deaths and deaths in the first 24h omitted.

Table 2: Estimation of predictive models derived from the primary model including predictors associated to post-discharge death among 20506 observations of children less than 15 years old and 750 deaths in the first 90 days following discharge.

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* WHZ: weight for height. *SD: standard deviation. See definitions in Table S1. ^BIt refers both community deaths and deaths in a readmission during follow-up period. ᵠHR: Hazard ratios. HR and confidence intervals were derived from a Cox regression model .*Confidence intervals. [@] P-value was derived from Wald test. ^{*}History of current disease reported by the child carer. ⁿHospital deaths and deaths in the first 24h omitted. δ AUC: area under the curve.

Table 3: Model's characteristics at probability cut-offs ensuring sensitivity of >80%

ᶽAUC: area under the curve; *CI: confidence intervals. ᵑPPV: positive predictive value.

Supplementary information

Post-Discharge Mortality Prediction in Sub-Saharan Africa

Contents

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Table S4: Estimation of a predictive model including predictors associated to postdischarge death among 1655 observations of infants less than 3 months old and 94 deaths in the first 90 days following discharge……………………………………………………….12

Figure S1. Flow chart on how hospital readmissions of children were selected.

Figure S3: Time-varying area under the curve of predictive models estimated from primary model.

A) Comparison among primary model and estimated models 2 and 3. B) Comparison between training and validation set of estimated model 4 which includes exclusively infants <3 months.

Figure S4: Number of admissions due to specific diseases over time

Table S2: Socio-demographic, clinical characteristics and univariate analysis of predictors on admission associated to post-discharge mortality among 2055 observations of infants less than 3 months of age and 126 post-discharge deaths in southern Mozambique.

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'SD: standard deviation. *'WAZ*: Weigh-for-age. ^{β}It refers both community deaths and deaths in a readmission during follow-up period. ^{*f*}Percentage represents risk among children with same characteristics. ^aUnivariate model based on 2055 infants under 3 months and 126 deaths. ^UHR: Hazard ratios. Hazard ratios and confidence intervals were derived from a Cox regression model.*Confidence intervals. ⁰ P-value was derived from Wald test. "Validated predictive model base of 80% of data (1636 infants). BCS: Blantyre coma score. ^{*}History of current disease reported by the child carer. ^{*n*}Hospital deaths omitted.

Table S3: Socio-demographic and clinical characteristics of children <15 years admitted at MDH comparing training (80% of data) and validation set (20% of data).

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ᶧSD: standard deviation. ᶤWHZ: Weigh-for-height.See definitions in Table S1.ᶽWAZ: Weight-for-age. ᵝIt refers both communitary deaths and deaths in a readmission during follow-up period. ʱPercentage represents risk among children with same chacarteristics. ᵠHazard ratios and confidence intervals were derived from a Cox regression model.*Confidence intervals. ⁶ P-value was derived from Wald test. BCS: Blantyre coma score. *History of current disease reported by the child carer. "Hospital **deaths ommited.**

Table S4: Estimation of a predictive model including predictors associated to postdischarge death among 1655 observations of infants less than 3 months old and 94 deaths in the first 90 days following discharge.

*SD: standard deviation. *'WAZ*: Weigh-for-age. ^Blt refers both community deaths and deaths in a readmission during follow-up period. Ψ HR: Hazard ratios. HR and confidence intervals were derived from a Cox regression model.*Confidence intervals. ⁰ P-value was derived from Wald test. BCS: Blantyre coma score. ^{*}History of current disease reported by the child carer. ⁿHospital deaths omitted.

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