- ¹ Physiologically driven, altitude-adaptive model for the
- ² interpretation of pediatric oxygen saturation at altitudes above
- 3 2000 m a.s.l.
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21 ABSTRACT

22 Measuring peripheral oxygen saturation (SpO_2) with pulse oximeters at the point of care is widely established. 23 However, since SpO₂ is dependent on ambient atmospheric pressure, the distribution of SpO₂ values in populations 24 living above 2000 m a.s.l. is largely unknown. Here, we propose and evaluate a computer model to predict SpO₂ 25 values for pediatric permanent residents living between 0 and 4000 m a.s.l. Based on a sensitivity analysis of 26 oxygen transport parameters, we created an altitude-adaptive SpO_2 model that takes physiological adaptation of 27 permanent residents into account. From this model, we derived an altitude-adaptive abnormal SpO₂ threshold 28 using patient parameters from literature. We compared the obtained model and threshold against a previously 29 proposed threshold derived statistically from data and two empirical datasets independently recorded from 30 Peruvian children living at altitudes up to 4100 m a.s.l. Our model followed the trends of empirical data, with the 31 empirical data having a narrower healthy SpO₂ range below 2000 m a.s.l., but the medians did never differ more 32 than 2.29% across all altitudes. Our threshold estimated abnormal SpO₂ in only 17 out of 5981 (0.3%) healthy 33 recordings, whereas the statistical threshold returned 95 (1.6%) recordings outside the healthy range. The strength 34 of our parametrised model is that it is rooted in physiology-derived equations and enables customisation. 35 Furthermore, as it provides a reference SpO₂, it could assist practitioners in interpreting SpO₂ values for diagnosis, 36 prognosis, and oxygen administration at higher altitudes.

37 New & Noteworthy

Our model describes the altitude-dependent decrease of SpO₂ in healthy pediatric residents based on physiological equations and can be adapted based on measureable clinical parameters. The proposed altitude-specific abnormal SpO₂ threshold might be more appropriate than rigid guidelines for administering oxygen that currently are only available for sea level patients. We see this as a starting point to discuss and adapt oxygen administration guidelines.

43 **Keywords:** Hypoxemia, Altitude, Oxygen Saturation, Model, Child Health, Pneumonia, Physiological Adaptation.

44 INTRODUCTION

45 Acute lower respiratory infections (ALRI) are a major health burden in low- and middle-income countries. 46 Childhood pneumonia accounts for 14% of all deaths in children worldwide under five years of age (45), of which 47 95 % occur in low resource settings (41). Common conditions observed in ALRI are dyspnoea and hypoxemia, an 48 abnormally low level of oxygen saturation in the arterial blood (SaO₂), that can lead to cyanosis and subsequently 49 to death (43). A rapid and non-invasive estimation of hypoxemia can be obtained through pulse oximetry that 50 measures peripheral oxygen saturation (SpO_2). Pulse oximetry has become a suitable technology for application in 51 low resource settings due to the simplicity of use in combination with mobile phones and non-invasiveness of the 52 device (20, 27). The use of pulse oximeters and supplemental oxygen in clinical applications at the point of care 53 has shown to drastically reduce death rates (8). However, in countries where these devices are needed most, 54 health personnel have only slowly started to gain access.

55 The interpretation of SpO₂ values for hypoxemia is challenging, especially for health personnel not familiar with 56 respiratory physiology and measurement principles of pulse oximeters. The World Health Organization (WHO) 57 recommends the administration of oxygen when SpO_2 drops below or is equal to 90% (44). This fixed threshold 58 oversimplifies hypoxemia treatment (7). It does not provide an indication on when to stop treatment and does not 59 permit adaptation to the local conditions. Namely, in many rural areas, oxygen is a scarce and precious resource 60 and therefore only restrictively administered. Altitude has a direct influence on SpO₂ as the air pressure decreases, 61 and consequently, the alveolar oxygen partial pressure decreases with increasing altitude (43). Thus, the treatment 62 of ALRI (i.e. administration of oxygen), diagnosis and prognosis, might be affected at higher altitudes and the 63 recommended oxygen administration guidelines at sea level may not be applicable. However, before determining 64 treatment thresholds at higher altitudes, healthy values in this environment need to be established.

In this work, we introduce an altitude-adaptive SpO₂ model and propose a model-derived altitude-adaptive abnormal SpO₂ threshold. The physiology-backed altitude-adaptive model describes SpO₂ values of healthy children living permanently at altitudes up to 4000 m a.s.l. With this model, we aim to provide a better understanding of healthy SpO₂ values at altitudes above 2000 m a.s.l. for healthy children. The altitude-adaptive abnormal SpO₂ threshold is obtained by setting the model parameters to abnormal values found in hypoxemic patients. We evaluate these results with a novel dataset obtained from healthy children living in the rural Andes of Peru.

72 Related Work

73 The current literature presents two modelling approaches that describe the relationship between SpO_2 and 74 altitude.

75 Subhi et al. developed a statistical model of the SpO₂ distribution across altitudes that is based on empirical 76 observations from healthy children, and derived an altitude-adaptive threshold for hypoxemia from this model 77 (40). Data were obtained through a literature review of studies performed between 0 and 4018 m a.s.l. A linear 78 random effects meta-regression was performed to predict mean and 2.5th centile SpO₂ with an exponential 79 equation. This 2.5th centile of healthy children's SpO₂ at each altitude was proposed as an altitude-adjusted 80 hypoxemia threshold. It is unclear why this specific, statistically derived threshold was chosen. The obtained 81 statistical model and threshold also did not take other influencing factors, such as measurement protocols, choice 82 of oximeter technology, ethnicity and age range of the studied subjects, into account.

83 Our group developed a computer model that described the pathway of oxygen throughout the cardio-respiratory 84 body compartments (24, 25). It implemented the oxygen cascade described by West (43). The model used well-85 established physiological equations to explain how the partial oxygen pressure and oxygen concentrations are 86 interrelated between alveolar gas and peripheral blood (24) (Figure 1). The oxygen cascade describes the oxygen 87 loss from the partial pressure of inspired air to the resulting measurements of SpO₂ by a pulse oximeter. Therefore, 88 the model was based on physiological parameters and integrated pulse oximeter measurement inaccuracies as 89 reported by the manufacturer. A shortcoming of the model was that it assumed many physiological parameters to 90 be constant and therefore did not consider altitude adaptation. Consequently, it could not correctly describe SpO₂ 91 measured at higher altitudes, especially in people adapted to these conditions such as permanent residents.

In a recent prospective study, Rojas-Camayo et al. recorded SpO₂ from 6289 subjects ranging from infants to elderly
people in the Peruvian Andes at 15 altitudes from 154 m to 5100 m a.s.l. (36). They reported the 2.5th, 10th, 25th,
50th, 75th, 90th and 97.5th centile of the empirical data. This data has not been used to derive a hypoxemia threshold
thus far.

96 MODELLING

97 Altitude-adaptive SpO₂ model

98 Starting from the previously established computer model of the oxygen cascade (24), we modified this model to 99 include physiological adaptation to high altitudes. We adjusted parameters that had been found to change with 100 altitude in permanent residents (see Table 1 for an overview of all parameters used). Briefly, the existing model of 101 the oxygen cascade described the pathway of oxygen throughout the cardio-respiratory body compartments 102 (Figure 1) by using physiological equations (see appendix). The model was originally developed to estimate the 103 "virtual shunt" (VS) describing the overall loss of oxygen content between the alveolar gas and arterial blood 104 compartments (2), with SpO_2 and inspired oxygen (FiO₂) values as input parameters. An increase in the VS is one 105 of the main causes of hypoxemia (43).

The above mentioned oxygen cascade model, originally developed for adults, can be adapted to a pediatric model as there are no indications that the underlying physics of gas exchange are any different in children (28). We identified relationships between altitude adaptation and parameters of the oxygen cascade, such as atmospheric pressure, haemoglobin concentration (cHb), alveolar partial pressure of carbon dioxide (pACO₂), and the respiratory quotient (RQ). In addition, we devided VS into two components (Figure 1): 1) incomplete capillary diffusion (diffusion defect between the alveolar and capillary , VS_{diff}) and 2) incomplete perfusion with intrapulmonary shunt (perfusion defect, VS_{perf}).

113 We made the following assumptions for the model of a healthy subject: there is no oxygen loss between the alveoli 114 and the end-capillaries (no incomplete capillary diffusion, VS_{diff}=0) and SpO₂ is equal to SaO₂ (24). These assumptions had the following consequences: the alveolar oxygen partial pressure (pAO₂) is equal to the partial 115 116 pressure of oxygen in the end capillaries , the alveolar oxygen saturation is equal to the end capillaries oxygen 117 saturation, and the oxygen content in the alveoli is the same as the in the end capillaries. For the parameters cHb 118 and RQ, we extracted the healthy values at two altitudes (0 m and 4600 m a.s.l.) from the literature (31, 32, 43) 119 and linearly interpolated the parameters between these two altitudes. A linear interpolation was chosen because 120 a sensitivity analysis revealed only small changes upon variation of these parameters (see appendix). For high 121 altitudes (i.e. 4600 m a.s.l.), pACO₂ was derived from an interpolation of values reported by Rahn and Otis (35), as 122 well as de Meer (32) because the literature presented less coherent values; while for sea level, direct values from

123 Marcdante (30) and West (43) were used. With this information, the oxygen cascade enabled us to estimate the 124 expected SpO_2 range at a specific altitude. Furthermore, we incorporated the technical tolerances that accounted 125 for the accuracy of pulse oximeters (i.e. $\pm 2\%$) determined according to device standards (21) into the model, as 126 shown in Karlen et al. (24). The pulse oximeter accuracy is an important component that is frequently neglected 127 by health practitioners, but influences the pulse oximeter readings and therefore diagnostic results. We include 128 this uncertainty in our model as we strive to better describe the physiology of lung function at different altitudes. 129 Therefore, in the following, when we mention the "healthy ranges", we refer to the physiological ranges obtained 130 by modelling SpO₂ based on minimum and maximum literature values of the physiological parameters, combined 131 with the pulse oximeter inaccuracies.

132 Altitude-adaptive abnormal SpO₂ threshold

133 Analogously, we derive an altitude-dependent threshold for abnormal SpO₂ by setting model parameters to 134 hypoxemia levels. Hypoxemia is defined as a reduced arterial partial pressure of oxygen (paO₂), which results in a 135 decrease of SpO_2 and increase of VS (43). At sea level, as reported in literature, we consider a patient to have 136 hypoxemia if the paO₂ level is below 80 mmHg (3, 26) and therefore SpO₂ decreases below 95%. Additionally, we 137 assumed that VSperf increases to above 5%. patients (31). From these assumptions, we recursively derived a disease 138 related increase of VSdiff of 19% at sea level. For higher altitudes, we were unable to retrieve any data from the 139 literature that would describe changes (increase or decrease) in VS (VSdiff or VSperf) or a numerical value for paO2 140 or pAO₂ under hypoxemia. Therefore, we assumed that the VS components remain constant across altitudes, and 141 the values for cHb, pACO₂, and RQ are similar in healthy and hypoxemic conditions.

142 MATERIALS AND METHODS

To assess the performance and plausibility of our novel altitude-adaptive SpO₂ model and threshold, we retrospectively evaluated them against a prospectively collected dataset, a previously published dataset, and another, statistical model with threshold.

146 Study design and data collection

Our data collection was embedded within a randomised controlled trial by the Swiss-Peruvian Health Research
 Platform set in the Cajamarca region in the northern highlands of Peru, located in the provinces of San Marcos and

Cajabamba. Our study harnessed the operational and logistical setup of this trial, which assessed the efficacy of an
 Integrated Home-environmental Intervention Package (IHIP-2) to improve child respiratory, enteric, and early
 development outcomes (19).

152 The trial was approved by the Universidad Peruana Cayetano Heredia ethical review board and the Cajamarca 153 Regional Health Authority. The trial was registered on the ISRCTN registry (ISRCTN26548981). A total of 317 154 children aged between 6 and 36 months were enrolled, and informed written consent was obtained from the 155 children's guardians. A total of 9 field workers (FWs) were trained to visit the children on seven fixed geographical 156 routes. Children were preassigned to these routes and visited in parallel by FWs to perform a mobile health 157 assessment once a week over the course of 60 weeks (6 weeks pilot, followed by a 54-week trial from February 158 2016 to May 2017, excluding 4 weeks of public holidays). FWs had experience from earlier research projects in 159 collecting basic vital signs and symptoms (17, 18), received five additional days of educational training for the 160 collection of morbidity data, and underwent one month of practical training before the study started (pilot). FWs 161 were equipped with a TAB 2 A7-10 tablet (Lenovo Group Ltd, Beijing, CN). The tablet had a custom mHealth app 162 installed that was developed using the lambdanative framework (34). It recorded a photoplethysmogram (PPG) 163 using an USB connected CE marked iSpO₂ Rx pulse oximeter (Masimo International, Neuchatel, CH) with a multisite 164 Y-probe, and derived SpO₂ and heart rate (HR). FWs placed the probe on the child's thumb, index finger, or sole of 165 the foot for the measurement of PPG, HR, and SpO₂. Simultaneously, respiratory rate was recorded using the RRate 166 app module (22). In addition, the app acquired location and altitude using the embedded global positioning system 167 (GPS) sensor. Furthermore, the app metadata regarding the visit and the recordings such as child ID, timestamps, 168 and child agitation during the vital signs measurements were acquired. All electronically collected data was 169 uploaded from the app into a digital research database (16). Health seeking behaviour and other relevant 170 endpoints were reported in a paper-based, validated questionnaire (18), quality checked, and digitised at the end 171 of the study.

172 Post processing

The IHIP-2 vital signs data obtained from the pulse oximeter were post processed to guarantee high data quality.
The PPG, SpO₂, HR, and perfusion index (PI, indication of signal strength) time series from the main trial period
were imported into Matlab (R2017b, MathWorks Inc., Natick, USA) where a signal quality index (SQI) for the PPG
was calculated (23). We segmented the recordings into segments with SQI > 45. Segments with lower quality (SQI)

177 \leq 45) and with no computed SpO₂ were excluded. Furthermore, entire recordings were excluded if a single segment 178 duration was shorter than 12 s or the combined length of remaining segments was shorter than 15 s, the range 179 (5th - 95th centile) of SpO₂ exceeded 5%, and the HR range surpassed 20 bpm in combination with a low perfusion 180 (mean PI \leq 0.8). We also excluded SpO₂ values below 60% as they are rare and typically associated with severe 181 clinical cyanosis (46), which was clearly absent in the IHIP-2 cohort. These values also fall in a range where the 182 performance of the pulse oximeters used were not specified by the manufacturer (70% to 100%). Additionally, as 183 each child was always scheduled to be measured weekly at the same altitude (i.e. at home), we verified the 184 consistency of the altitude provided by the GPS. We excluded recordings that contained no altitude information, 185 and altitude outliers that were more than three scaled median absolute deviations away from the median altitude 186 of each child. Altitude outliers could have occurred because at home measurements were not always possible, and 187 because GPS altitude estimates were dependent on weather, the number of available satellites, and other factors. 188 Finally, we excluded measurements which were recorded following a healthcare center visit or the presence of 189 cardio-respiratory or diarrheal disease symptoms in the week preceding the recording. For each remaining high 190 quality recording, we reported the median SpO₂ over the combined segments of a measurement and the median 191 altitude per child, which was then used for the analysis.

192 Evaluation

193 Model

194 To compare our model with the available datasets, we visualised the altitude dependence of SpO₂. We applied a 195 locally weighted scatterplot smoothing (lowess) function (5) to all SpO₂-altitude data pairs collected during the 196 IHIP-2 trial. We limited the comparison to the range of available data (2000–4000 m a.s.l.) to avoid extrapolation 197 errors. Instead of the LMS method used by Rojas-Camayo et al. (36), we reported the centiles of their data with a 198 lowess smoother to ensure equivalent processing of both datasets. Furthermore, we computed the deviations 199 from interpolated medians of both empirical data sets to the model median for each altitude expressed as percent 200 of the respective model value and reported the mean, minimum and maximum deviations. Additionally, we 201 calculated the absolute range of SpO₂ values at each altitude for both the model and the empirical data sets and 202 reported mean, minimum and maximum range.

203 Threshold

To visualise the differences between the hypoxemia/abnormal SpO₂ thresholds and oxygen administration guidelines that have been proposed, we graphically compared the altitude-adaptive abnormal SpO₂ threshold, the

statistical hypoxemia threshold, and the WHO guideline for oxygen administration (90%) with the 2.5th centile (lowess smoothed) data of children 1 to 5 years old reported by Rojas-Camayo et al. (36). We further computed the number of measurements in the healthy IHIP-2 data that would have been wrongly classified as abnormal (false positives) when using either the altitude-adaptive abnormal SpO₂ threshold or the statistical hypoxemia threshold. The false positives are children that are healthy, but likely would receive additional medical attention due to the low SpO₂ reading.

212 **RESULTS**

213 We obtained an altitude-adaptive computer model to describe the expected SpO₂ range in healthy children at 214 higher altitudes, and based on this model proposed a threshold for an abnormal range that could indicate 215 hypoxemia. The parameters used in the mathematical description of the model to define healthy and abnormal 216 ranges are available in Table 1. Out of the 12634 SpO₂ measurements obtained from 310 children over the course 217 of a year, we retained 5981 measurements from 297 children that were considered complete (contained both GPS 218 and PPG data), featured good quality PPG data, reasonable SpO₂ (> 60 %) and were recorded when no respiratory 219 disease symptoms or other health issues were reported (410 recordings). At the study start, the mean age of the 220 children was 20.5 months (SD 6.2 months, range: 6-36 months). Each child contributed to a mean of 20.1 (SD 9) 221 repeated measurements. Twenty-one children lived above 3000 m a.s.l. and 8 above 3500 m a.s.l. (Table 2). 222 Therefore, a total of 392 (6.6%) measurements above 3000 m a.s.l. were available.

223 Model

224 Our altitude-adaptive model provided a SpO_2 of 97.4% at sea level with a healthy range between 93.5% and 100% 225 SpO₂ (Figure 2 and Table 3, high resolution data including model available at (11)). The SpO₂ of the model decreased 226 with increasing altitude to 89.6% at 4000 m a.s.l. with a healthy SpO_2 range from 82.3% to 94.1%. The 2.5th and 227 97.5th centiles reported by Rojas-Camayo et al. largely followed the same trend as those acquired in the IHIP-2 228 trial, but had a smaller absolute range (Figure 2). Up to 3800 m a.s.l., the 2.5th centiles of both empirical data sets 229 were entirely within the lower boundary of the altitude-adaptive SpO₂ model's proposed healthy range, whereas 230 at higher altitudes above 3800 m a.s.l., the 2.5th centile of the IHIP-2 data slightlyfell below this lower boundary. 231 The upper boundary of the altitude-apdaptive SpO₂ model's healthy range followed the IHIP-2 data 97.5th centile 232 closely, while it was slightly exceeded by the 97.5th centile data from Rojas-Camayo et al. between 1500 and 233 3100 m a.s.l by up 0.5 %. In particular, the model showed absolute ranges very similar to both empiricial lowess

filtered data sets (model: mean absolute SpO₂ range: 8.66%, min: 6.42%, max: 11.78%; IHIP-2: mean absolute SpO₂
range: 8.75%, min: 6.75%, max: 11.22%; Rojas-Camayo: mean absolute SpO₂ range: 5.53%, min: 3.43%, max:
8.92%). Furthermore, the model differed very little from the interpolated median of the empirical data sets (IHIP2, deviation of model in percent: mean deviation: 1.5%, min: 0.01%, max: 2.29%; Rojas-Camayo: mean deviation:
1.51%, min: 0.02%, max: 1.95%).

239 Threshold

The altitude-adaptive abnormal SpO₂ threshold followed a similar pattern as the 2.5th centile of Rojas-Camayo's empirical data with 88.8% vs 94% at 2000 m a.s.l. and 80.1% vs 83.8% at 4000 m a.s.l. (Figure 3, see also Table 3). The 2.5th centile threshold explored by Subhi et al. had an SpO₂ of 92.8% at 2000 m a.s.l. and then rapidly diverged towards much lower SpO₂ values for higher altitudes (75.4% at 4000 m a.s.l.). When comparing the two thresholds and their performance for our empirical dataset, the altitude-adaptive threshold estimated abnormal SpO₂ in only 17 out of 5981 (0.3%) healthy recordings, whereas the 2.5th centile threshold explored by Subhi et al. returned 95 (1.6%) false positives.

247 DISCUSSION

248 We proposed an altitude-adaptive model that estimates a healthy SpO₂ range for children living permanently at 249 altitude and have shown that this proposed healthy SpO₂ range matches empirical data recorded from a pediatric 250 population living in the Andes. From this model, we derived an altitude-adaptive threshold for abnormal SpO₂ 251 values. The diagnosis of pneumonia and other respiratory diseases is challenging at altitude, as the most common 252 diagnostic criteria, such as the respiratory rate and oxygen saturation, are dependent on altitude. Our work 253 contributes towards making the management of childhood pneumonia, one of the major causes of child mortality 254 in low resource settings, more objective by attempting to better describe healthy changes of respiratory physiology 255 found in adapted residents. Equipping health workers with mobile pulse oximeters has become an affordable 256 solution, is being evaluated at a large scale (29), and has potential for improving pneumonia treatment at a 257 reasonable cost (12). However, the measurement and interpretation of SpO_2 can be complicated. Computerised 258 assistance and interpretation of the measurements could ensure reliability of these measurements and provide a 259 meaningful decision support tool to health workers at the central and peripheral level. The proposed adaptive, 260 physiology-based model could provide a basis for the necessary computations because it provides a reference for 261 healthy values at higher altitudes.

Our model is unique as the adjustment of the parameters can be tuned individually, based either on measurements or on known parameter ranges, and it is based on physiology. It was developed considering, where available, literature-based physiological parameter values of Peruvian Andes residents that are adapted to this environment. These parameters could be adjusted without altering the underlying model for other populations with known differences in genetic or physiological adaptation mechanisms (e.g. Himalayan residents) (1).

267 In contrast to our parameterized model, Subhi and colleagues fitted empirical data collected from across the world 268 into a statistical model describing the SpO₂ distribution using centiles (40). The statistical model was built using 269 aggregated data collected from mixed populations using pulse oximeters with partially unknown specifications. 270 The statistical model therefore cannot be adjusted to factors such as population-specific variations or varying 271 technical specifications (e.g. differing accuracy of pulse oximeter brands or types). In relation to the two empirical 272 data sets mentioned in this publication, and in comparison to our proposed abnormal SpO₂ threshold, the 273 statistical threshold provided a very sensitive cut-off at lower altitudes (up to 3300 m a.s.l.). However, it 274 underestimates potentially abnormal SpO₂ values at higher altitudes. Most likely, this underestimation of the 275 abnormal SpO₂ values at higher altitudes is due to less data samples being available for the statistical modeling. 276 Our physiological model was not affected by data sparsity, which is a distinctive feature and clear advantage at 277 higher altitudes. Both model thresholds, and the studied data sets, supported the current WHO constant threshold 278 of 90% SpO₂ for oxygen administration at altitudes below 1500 m a.s.l.

The altitude-adaptive model described the SpO₂ ranges observed from the empirical data sets with highly similar mean absolute ranges., However, the two empirical datasets presented in this work originate solely from the Peruvian Andes and a single type of pulse oximeter. To further validate the model, it will be crucial to apply data from other regions and ethnicities, and establish if a customised model is required when used in different parts of the world. Such data collection should be accompanied by a gold standard, such as blood gas measurements with information on cHb, SaO₂, paO₂ and paCO₂, in order to pinpoint the exact sources of potentially observed differences.

At higher altitudes above 3800 m a.s.l., we notice higher deviations in the model compared to what is seen in the empirical data due to a slower decline of SpO₂ in the model. We suspect that this is directly linked to the assumptions we made during the modelling of healthy ranges. We assumed that cHb and RQ change linearly with altitude. However, the adaptation process is likely more pronounced at higher altitudes (6) and might contribute to non-linear parameter changes.

291 Our assumptions to define the abnormal physiological parameters could limit the validity of the abnormal 292 threshold. We only based our assumptions on literature values that referred to sea level patients. Due to the 293 underlying changes in physiology caused by adaptation, disease manifestation and progression, symptoms could 294 be different at high altitudes compared to at sea level. Furthermore, it is unclear if comorbidities that have not 295 been captured in the present modelling, such as malnutrition, iron deficiency, or diarrheal diseases that are known 296 to negatively influence outcomes of patients with pneumonia (4, 37, 39), would also influence the model 297 parameters. Additional empirical data of sick children are needed to establish models that describe the 298 dependence of these parameters to altitude. For example, anaemic children display altered ranges for blood gas 299 parameters and their actual health status is not entirely captured through our cardio-respiratory model based on 300 SpO_2 measurements. SpO_2 and derived hypoxemia estimations reflect only the proportion of O_2 that is bound to 301 Hb and not the total O₂ carrying capacity and concentration. Consequently, pulse oximeter assessments are blind 302 to the effective O_2 available in the tissues. Also, cardiac output, an alternative path to modulate O_2 delivery (14), 303 is not easily obtainable with pulse oximetry alone. Thus, clinicians need to take the overall clinical situation of the 304 child into consideration and evaluate treatment options accordingly when interpreting hypoxemia thresholds (10).

305 To assess the performance of the model, we limited the comparison to altitudes from 2000 to 4000 m a.s.l. where 306 corresponding empirical data was available. The data contained weekly measurements for each child repeated 307 over a full year (mean: 20.1, SD: 9), therefore representing the expected measurement and physiological variability 308 within a healthy subject. Among the children recruited from the Cajamarca region during the IHIP-2 trial, only 21 309 lived above 3000 m a.s.l. which increases the variability in the data. Nevertheless, we observed very similar SpO_2 310 ranges from Rojas-Camayo et al. (36). Despite the high numbers of repeated measurements and rigid 311 measurement protocols, both datasets showed a high variability in the measured SpO2. For example, in the IHIP-2 312 dataset, at 2000 m a.s.l. a healthy range corresponded to 11% (Table 2). Our model represented this large range 313 of possible healthy values accurately. Nevertheless, the inter- and intra-individual variability could originate from 314 a number of sources not incorporated in the model. Circadian variation in pediatric SpO₂ has been reported (42) 315 and we did not account for such daytime differences. Furthermore, there are known sex differences in adults (1), 316 which could also apply to the pediatric population. Although we used the most recent pulse oximeter technology 317 and performed continuous measurements for at least a minute with a rigorous approach to PPG post-processing 318 for high quality, not all the sources for measurement errors in pulse oximetry, such as poor perfusion, inacurate 319 probe positioning, or ambient light interference (13), could be fully excluded in this dataset.

Additionally, it is important to note that neonates were not considered in the modeling process. Neonatal blood is known to benefit from the high affinity of fetal haemoglobin and would have changed the oxygen dissociation curve considerably (33). Since hyperoxia in neonates leads to oxidative stress with potentially severe health complications (15), the definition of an abnormal threshold and consequently the guideline for oxygen administration would require a more detailed, separate discussion for this population.

325 We established an altitude-adaptive abnormal SpO₂ threshold based on physiologically plausible values. Our 326 results show that using such a threshold is most relevant at altitudes above 2000 m a.s.l. The 90% SpO₂ threshold 327 recommended by the WHO for oxygen administration in patients living at sea level clearly does not apply to these 328 altitudes. Compared to the previously published statistical altitude-dependent threshold by Subhi et al. (40), our 329 threshold leads to fewer detections of false positives (healthy children falsely categorized as hypoxemic). 330 Conversely, while Subhi et al. also promoted the use of an altitude-dependent threshold at higher altitudes 331 (2500 m a.s.l.), their threshold is very conservative at altitudes below 2950 m a.s.l. but more lenient at higher 332 altitudes, where it decreases very steeply which might exclude a number of patients in need of supplemental 333 oxygen.

334 Outlook

Thus far, experts have not agreed on a definition for abnormal SpO₂ thresholds at altitudes higher than sea level. To date, no reliable SpO₂ data from children suffering from hypoxemia and ALRI at altitude are available. The advancement of research for developing better tools to diagnose pneumonia and ALRI at altitude would greatly benefit from access to publicly available, comprehensive data sets obtained from sick children.

With pulse oximeters increasingly being used as monitors for ALRI diagnosis and treatment, additional research is
 urgently needed to provide a reliable description of the SpO₂ distribution at altitude, and to develop guidelines of
 oxygen administration for hypoxemic children living in these settings.

Furthermore, knowledge of abnormal SpO₂ values at high altitudes could help in the development of new decision support tools for health workers operating in low resource settings with the goal to improve clinical management of hypoxemia in children with ALRI in the future.

345 CONCLUSION

346 Improvement of SpO₂-altitude models present a first step towards an integration of pulse oximetry in low resource 347 settings and could further the development of valid altitude-dependent thresholds for treatment of childhood 348 pneumonia and other ALRI. We developed an altitude-adaptive physiology-backed SpO₂ model using an existing 349 physiological model using the concept of VS adjusted for published ranges of values for pACO₂, cHb, and RQ. Based 350 on this model, healthy ranges and an altitude-dependent abnormal SpO₂ threshold are suggested that are based 351 on physiological variations of vital parameters. With the increased availability of sensors and digitalised systems in 352 low resource settings, parametrised models could provide additional valuable support to primary health workers 353 to understand the patient's condition at the point of care, and choosing treatment options based on objectively 354 obtained physiological measurements.

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366 **COMPETING INTERESTS**

367 The authors declare no competing interests.

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487		

488 TABLES

- 489 Table 1: Physiological parameters obtained from the literature used to implement the altitude-adaptive SpO₂ model. The healthy
- 490 ranges (min, max and mean) describe known and expected values in a healthy subject. Parameters that are expected to change
- 491 under hypoxemic conditions are reported in the last column. We do not differentiate between adults and children. *same value
- 492 range (min max) assumed as at sea level. pACO₂: alveolar partial pressure of carbon dioxide, RQ: respiratory quotient, cHb:
- 493 haemoglobin concentration, VS_{perf}: perfusion defect, VS_{diff}: diffusion defect, pAO₂: alveolar partial pressure of oxygen, FiO₂:
- 494 fraction of inspired O₂, P_{atm}: atmospheric pressure, PH₂O: vapour pressure of water.

			Healthy ranges						Hypoxemia
Parameter	Unit	Ref.		Sea level 4500 m a.s.l.					
			<u>min</u>	<u>mean</u>	<u>max</u>	<u>min</u>	<u>mean</u>	max	
pACO ₂	mmHg	(30, 32, 35)	35	40	45	23	28.3	33	Assumed to be
RQ	unitless	(32, 43)	0.8	0.8	1	1	1	1	equal to healthy
cHb	g/100 ml	(43)	12	15	17.5	17	20	22.5	range
VSperf	%	(31)	0	2	5	0	2	5	≥ 5*
VS _{diff}	%	Derived	0	0	0	0	0	0	≥ 19*
pAO ₂	mmHg	Derived	Derived by the alveolar gas equation, with parameters FiO ₂ , P _{atm} , PH ₂ O, pACO ₂ and RQ						

495

- 497 Table 2: IHIP-2 pulse oximeter data: Distribution of children, number of measurements and SpO₂ per altitude. Age range of the
- 498 children at study start: 6-36 months (mean: 20.5 months, SD: 6.2 months), 21 children above 3000 m a.s.l. (392 measurements,
- 499 6.6 % of total number of measurements), 8 above 3500 m a.s.l. (171 measurements, 2.9 % of total number of measurements),
- 500 mean number of measurement per child: 20.1 (SD 9).

Altitude (m a.s.l.)	No. children	Age at study start (months)		No. of measurements	No. of measurements per child			SpO₂ (%)		
		Mean	SD		Min	Mean	Max	Min	Median	Max
2000	78	21.0	6.0	1668	1	21.4	39	88.9	97.2	100.0
2100	37	20.4	6.1	803	1	21.7	43	87.4	97.2	100.0
2200	6	15.7	5.0	109	7	18.2	24	88.6	96.6	99.7
2300	15	19.5	6.7	389	5	25.9	34	89.2	96.7	100.0
2400	3	24.0	2.0	77	19	25.7	31	91.9	96.2	100.0
2500	13	20.5	5.7	264	4	20.3	33	87.1	96.0	100.0
2600	22	21.6	5.6	472	2	21.5	39	81.1	96.0	100.0
2700	39	21.5	6.8	753	4	19.3	32	85.2	95.6	100.0
2800	20	21.8	5.7	316	1	15.8	34	83.9	95.2	99.5
2900	18	17.7	6.0	276	1	15.3	27	85.6	94.7	98.6
3000	25	20.2	7.2	462	5	18.5	37	81.5	94.5	100.0
3100	8	20.0	6.5	111	1	13.9	24	80.7	94.4	100.0
3200	3	19.3	6.0	76	12	25.3	38	85.7	93.4	98.2
3300	1	20.0	0.0	19	19	19.0	19	89.9	92.4	95.7
3400	1	17.0	0.0	15	15	15.0	15	86.0	92.9	95.4
3500	0	-	-	0	-	-	-	-	-	-
3600	5	19.4	4.7	107	9	21.4	30	84.3	92.7	98.7
3700	1	28.0	0.0	24	24	24.0	24	86.4	92.5	95.6
3800	1	23.0	0.0	22	22	22.0	22	80.8	87.8	93.9
3900	1	14.0	0.0	18	18	18.0	18	81.6	88.4	90.6
total	297	20.5	6.2	5981	-	20.1	-	-	-	-

502 Table 3: Values of the healthy ranges of the altitude-adaptive SpO_2 model including its median and the abnormal SpO_2

503 threshold, per altitude. More granular altitude steps available in (11).

AltitudeUpper healthy range(m a.s.l.)(% SpO2)		Lower healthy range (% SpO ₂)	Model Mean (% SpO₂)	Threshold (% SpO₂)	
0	100.0	93.6	97.4	92.9	
500	100.0	93.0	97.0	92.2	
1000	99.6	92.2	96.4	91.2	
1500	99.1	91.3	95.8	90.1	
2000	98.5	90.1	95.0	88.8	
2500	97.7	88.7	94.0	87.1	
3000	96.7	87.0	92.8	85.2	
3500	95.6	84.9	91.4	82.9	
4000	94.1	82.3	89.6	80.1	

504

505 FIGURES

506 Figure 1: Oxygen cascade describing the loss in oxygen partial pressure (pO_2) between inspired air and peripheral blood

507 measured with a pulse oximeter. The lines illustrate the standard situation for a healthy subject at sea level (continuous) and

508 at 4500 m a.s.l. (dashed). Virtual shunt describes the combined loss in oxygen content due to incomplete diffusion or perfusion

509 between alveolar and arterial compartments. Adapted from (24, 43).

510

Figure 2: The proposed altitude-adaptive SpO₂ model provides a healthy SpO₂ range (light grey area). The black dotted line
indicates the median SpO₂ estimated by the model. The parameters for the min, max and mean model parameters are given in
Table 1. The 2.5th-97.5thcentiles of the SpO₂ data from Rojas-Camayo et al. (light grey dashed lines) (36) and the Integrated
Home-environmental Intervention Package (IHIP-2) data set (black dashed-doted lines) that were both recorded in the Peruvian

515 Andes mostly fall into our proposed healthy range. The reported number of measurements per children for the IHIP-2 data can

516 be found in Table 2.

517

- $518 \qquad \textit{Figure 3: Comparison of proposed abnormal SpO_2 thresholds that would lead to oxygen administration in patients and existing}$
- 519 guidelines. The altitude-adaptive abnormal SpO₂ threshold (continuous black line) is based on the physiological model derived
- 520 in this work where a virtual shunt was applied. The threshold from Subhi et al.(40) is the 2.5th centile derived from
- 521 observations in healthy children collected in a literature review (dotted light grey line), and the 2.5th centile from Rojas-
- 522 Camayo et al.(36) is derived from a prospectively collected healthy pediatric sample in the Peruvian Andes (dashed grey line).
- 523 The WHO 90% oxygen administration guideline is a result of a working group consensus (starred grey line) (44) that is in use at
- 524 *lower altitudes.*

526 Appendix

527

Table A1: Parameters required for the calculation of the oxygen cascade with specification of dependency on other
 parameters and the expected change with increasing altitude.

Abbreviation	Unit	Name	Healthy adult value at 0 m a.s.l.	Dependency	Expected change with increasing altitude
pAO ₂	mmHg	Partial pressure of alveolar O ₂		FiO_2 , P_{atm} , PH_2O , $pACO_2$ and RQ	Decrease
<i>Sc</i> ′ <i>O</i> ₂	%	Oxygen saturation of end-capillary blood		pAO ₂	Decrease
<i>Cc'0</i> ₂	ml/100 ml	Oxygen content of end- capillary blood		BO ₂ , cHb, SAO ₂ , PAO ₂	Increase
CaO ₂	ml/100 ml	Oxygen content of arterial blood		BO ₂ , cHb, SaO ₂ , PaO ₂	Increase
PaO ₂	mmHg	Partial pressure of arterial O ₂		SaO ₂	Decrease
SaO ₂	%	Oxygen saturation of arterial blood			Decrease
P _{atm}	kPa	Ambient gas pressure	101.325		Decrease
FiO ₂	%	Fraction of inspired O ₂	21		No change
pACO ₂	mmHg	Partial pressure of alveolar CO ₂	40		Decrease
VSperf	%	Perfusion component of virtual shunt	2		No change
VS _{diff}	%	Diffusion component of virtual shunt	0		No change
RQ	unitless	Respiratory exchange ratio (O2 inspired/CO2 expired)	0.8		Increase
cHb	g/100 ml	Haemoglobin concentration in blood	12		Increase
PH ₂ O	mmHg	Vapour pressure of water	47		No change
BO ₂	ml O ₂ /(g Hb)	Oxygen-binding capacity of haemoglobin in blood	1.34		No change
$CaO_2 - CvO_2$	ml/100 ml	Arteriovenous oxygen difference (CvO ₂)	5		No change

532 **Equations**

533 For the entire computer model of the oxygen cascade, please consult (24, 25). See Table A1 for the 534 variable names.

535

536

537 Alveolar gas equation:

540

$$pAO_2 = FiO_2 * (P_{atm} - PH_2O) - \frac{pACO_2}{RQ} + pACO_2 * FiO_2 * \frac{1 - RQ}{RQ}$$

539 Severinghaus equation (38):

 $SAO_2 = \frac{1}{\frac{23400}{pAO_2^3 + 150 * pAO_2} + 1} * 100$

.

541 *O*₂ *Content equation:*

542
$$CxO_2 = BO_2 * \frac{cHb * SxO_2}{100} + 0.003 * pxO_2$$

Severinghaus-Ellis equation (9): 543

544
545

$$PaO_2 = (B + A)^{1/3} - (B - A)^{1/3}$$

 $A = \frac{11700}{\frac{1}{SaO_2 - 1}}$
 $B = \sqrt{50^3 + A^2}$

Virtual Shunt from perfusion defect (VS_{perf}): 546

547
$$VS_{perf} = \frac{CcO_2 - CaO_2}{CcO_2 - CaO_2 + (CaO_2 - CvO_2)}$$

548

Virtual Shunt from diffusion defect (VS_{diff}): 549

550 $pcO_2 = pAO_2 * (1 - VS_{diff})$

551

553 Sensitivity analysis

554 To display the influence of parameters on the output of the oxygen cascade, a sensitivity analysis was

- 555 performed (Figure A1). The parameters variation was chosen to reproduce the minimum and maximum
- value used in the altitude-adaptive SpO₂ model (Table 1). A change in pACO₂ had the highest effect,
- 557 followed by VS_{diff}, VS_{perf} and RQ. A change in cHb is negligible for the calculation of SpO₂, however, please
- note that it has a significant influence on availability of O₂ in the tissues.
- 559
- 560
- 561 Figure A1: Sensitivity analysis of the five main model parameters respiratory quotient (RQ) (top left), virtual
- shunt from diffusion defect (VS_{diff})(top right), virtual shunt from perfusion defect (VS_{perf}) (middle left), PACO₂
- 563 (middle right), and cHb (bottom left).







