

# Journal Pre-proof

Weight velocity in addition to latest weight does not improve the identification of wasting, or the prediction of stunting and mortality: a longitudinal analysis using data from Malawi, South Africa, and Pakistan

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32 **Abstract**

33 **Background:** In low/middle-income countries, most nutritional assessments use the latest  
34 weights, without reference to growth trajectory.

35 **Objective:** This study explores whether velocity, in addition to the latest weight, improves  
36 the prediction of wasting, stunting or mortality in the first two years of life.

37 **Methods:** We analysed a combined data set with weight and height data collected monthly  
38 in the first year of 3447 children from Pakistan, Malawi, South Africa, with height and  
39 survival recorded till 24 months. The main exposures were weight-for-age z-score (WAZ)  
40 at the end of each 2-month period and weight velocity-for-age z-score (WVZ<sub>2</sub>) across that  
41 period. The outcomes were wasting, stunting or all-cause mortality in the next 1-2 months.  
42 As a sensitivity analysis, we also used WVZ over 6 months (WVZ<sub>6</sub>), with matching WAZ.  
43 Cox proportional hazard models with repeated growth measures were used to study the  
44 association between exposures and mortality. Mixed Poisson models were used for stunting  
45 and wasting.

46 **Results:** Children who were already stunted or wasted were most likely to remain so.  
47 WVZ<sub>2</sub> was associated with a lower risk of subsequent stunting (RR 0.95; 95% CI 0.93-  
48 0.96), but added minimal prediction (difference in AUC = 0.004) compared to a model  
49 including only WAZ. Similarly, WVZ<sub>2</sub> was associated with wasting (RR 0.74; 95% CI  
50 0.72-0.76) but the prediction was only marginally greater than for WAZ (difference in  
51 AUC = 0.015). Compared to WAZ, WVZ<sub>6</sub> was less predictive for both wasting and  
52 stunting. Low WVZ<sub>6</sub> (but not WVZ<sub>2</sub>) was associated with increased mortality (HR 0.75,

53 95% CI 0.67-0.85), but added marginal only prediction to a model including WAZ alone  
54 (difference in C = 0.015).

55 **Conclusions:** The key anthropometric determinant of impending wasting, stunting, and  
56 mortality appears to be how far below the normal range the child's weight is, rather than  
57 how they reached that position.

58

59 **Keywords:** Children; Mortality; Growth Disorders; Stunting; Weight.

60

## 61 **Introduction**

62 Child undernutrition remains a major global health concern (1). World Health Organisation  
63 (WHO) definitions of malnutrition rely on single measures of weight and length/height and,  
64 more recently, mid upper arm circumference, as well as the combination of weight and  
65 length/height to give weight for height (WFH). The most common measure is that of  
66 weight, usually adjusted for age via a lookup table, plotting on a chart or converting to a  
67 standard deviation (Z) score. A weight for age Z score (WAZ)  $<-2$  is defined as moderate  
68 malnutrition and  $<-3$  as severe, while a length for age Z score (LAZ)  $<-2$  is defined as  
69 stunting (2). Regularly weighing aims to identify children who have lost or gained weight  
70 more slowly than their peers and are malnourished, so a measure that determines children  
71 with low weight velocity (WV) might be expected to be more discriminating than attained  
72 weight (AW) alone. In high-income countries, this has led to the concept of failure to thrive  
73 (FTT) or weight faltering, where the diagnosis depends on low WV based on two or more  
74 weights, at least while a child is within the normal range (3).

75 In contrast, in low- and middle-income countries (LMIC), decisions tend to be based only  
76 on the current weight, as interpretation of serial measurements requires more detailed  
77 record-keeping and WHO definitions of malnutrition include no velocity elements (4).

78 However, the WHO growth chart project has published age and sex-standardised norms for  
79 weight velocity over 4- and 8-week intervals (5) and they have argued that velocity  
80 measures are helpful in predicting later stunting (6). As severe malnutrition (SAM)  
81 becomes less common (7), the accurate diagnosis of moderate malnutrition (MAM)  
82 becomes more important, to target treatment to children who are most likely to benefit.  
83 Using only a low AW will also identify children who were born small but have grown

84 steadily since birth and supplementary feeding could even harm these children (8). Thus, a  
85 secondary screening measure of velocity could potentially be used to improve the  
86 specificity of diagnoses.

87 Few studies have examined whether velocity improves the prediction of adverse growth  
88 outcomes, compared to using a single weight (7, 8). Only one has considered outcomes  
89 beyond the age of 6 months or used increments greater than one month (9). Thus, it is  
90 unclear whether velocity adds value in addition to the latest weight alone, and if so, when,  
91 and over what time interval weight velocity might usefully be measured.

92 We thus aimed to use a large combined historical data set of longitudinal growth data, with  
93 a high prevalence of malnutrition, to examine the value of WV, in addition to AW, in  
94 identifying the onset of wasting and predicting stunting or mortality at different ages, and  
95 over differing time intervals.

## 96 **Methods**

97 This prospective study used data from three cohort studies conducted in low or middle  
98 income countries, all with monthly measurements collected for at least the first 12 months  
99 and outcome of interest up to 24 months. These were the Lungwena Child Survival Study  
100 (Malawi), the Africa Centre Vertical Transmission Study (South Africa), and the Lahore  
101 longitudinal study (Pakistan). Each study is briefly described below.

### 102 *Included cohort studies*

103 The Lungwena Child Survival Study was a cohort study of 795 women recruited between  
104 1995 and 1996 and their newborn children prospectively studied (10, 11). Children were

105 measured monthly to 18 months, then every three months up to 36 months. Children were  
106 measured at home by a research assistant using portable spring scales and length boards  
107 (12).

108 The Africa Centre Vertical Transmission Study study registered 2938 children (half HIV-  
109 positive) from seven rural, one semi-urban, and one urban primary health care clinic in  
110 KwaZulu-Natal, in a nonrandomised intervention cohort study between 2001 and 2005 (13,  
111 14). Weight and lengths were collected by research staff, using the WHO-recommended  
112 protocol, monthly until 12 months, then every three months to 24 months (12).

113 The Lahore longitudinal study enrolled infants born between 1984 and 1994, with 1314  
114 from a village area, 572 from a peri-urban slum, 921 from an urban slum, and 339 from a  
115 middle-class neighbourhood, and followed them monthly from birth up to 36 months (15,  
116 16). All infants were weighed and measured at home by specially trained research  
117 assistants, with the measuring technique checked monthly and instruments checked weekly  
118 (12, 17).

119 In the three studies, all deaths were recorded, but their causes were not recorded  
120 consistently between datasets. Sociodemographic information (age, sex, and country of  
121 origin) was self-reported. Moreover, HIV-positive mothers and subsequent HIV-positive  
122 children were excluded from the analyses.

### 123 *Statistical analyses*

124 The three datasets were combined into one database (excluding all HIV-positive mothers)  
125 and measurements were expressed as Z scores compared to the WHO growth standard (18).

126 Stunting was defined as length for age (LAZ)  $< -2SD$  and wasting as weight for length  
127 (WLZ)  $< -2$ . The analysis was then conducted per measurement rather than per child. For  
128 each attained weight, the exposures of interest was the Z score for that weight (WAZ) and  
129 the velocity for age (WVZ) across the period up to that attained weight, calculated using the  
130 WHO velocity standards(5) for both 2 and 6 month intervals. The outcome was whether or  
131 not the child was stunted or wasted at the next monthly measurement, or whether they had  
132 died. Attained WLZ and LAZ were treated as secondary or additional exposures.

133 Prediction models were constructed using weight velocity and the attained growth  
134 parameters at the end of the weight velocity period, to predict the next observation of  
135 wasting, stunting, or mortality. Since each growth measurement was used to predict the  
136 subsequent period's wasting and stunting risk, mixed Poisson models with repeated growth  
137 measures were used to study the associations. All observations were included in the model  
138 to maximise power while intra-personal correlations were captured as a random intercept.  
139 Results are reported as risk ratios (RR) with their respective 95% CI. Model predictive  
140 performance was assessed with the area under the receiver operating characteristic curve  
141 (AUC). The main analyses were adjusted for age at assessment, sex, and country of origin.  
142 In addition, analyses were stratified for incident and recurrent wasting/wasting for the  
143 predictors that had the highest AUC.

144 Cox proportional hazard models with repeated growth measures were used to study the  
145 association between exposures and mortality rate. The outcome variable was time to event  
146 (either death or censoring). The basic analyses were adjusted for age at assessment, sex,  
147 and country of origin (model 1). Additional models included the mutual adjustment among  
148 the growth parameters. Analyses are reported as hazard ratios (HR) with 95% confidence



149 intervals (CI). In the mortality analysis, the predictive ability was quantified using Harrell's  
150 C-index – which estimates the probability of concordance between observed and predicted  
151 responses (19).

152 Finally, sensitivity analyses were used to investigate whether the associations observed  
153 differed by age ( $\leq$  and  $>$  6 months). The interaction between binary age group ( $\leq$  vs.  $>$  6  
154 months) and weight variables were included in the corresponding regression models and  
155 can be interpreted as the ratio of RR (for wasting and stunting) and ratio of HR (for  
156 mortality). A ratio  $>$  1 indicates that the association of the weight variable with the outcome  
157 is stronger in the older ( $>$  6 months) group.

158 R 4.0.5 software was used to perform all analyses. A p-value  $\leq$  0.05 was considered  
159 statistically significant.

## 160 **Results**

161 The baseline characteristics of the included data are shown in Table 1. Of the 3447 children  
162 with any pairs of growth data, 98 children died, and in a third of these this followed a  
163 weight  $< -2SD$ . Of the included weights, the majority (78%) were within normal range, but  
164 20% were  $< -2$  (underweight) and only 1% were  $> 2$ . Using the first measurements available  
165 near to birth, 11.8% of children were  $< -2$  WAZ and 14.4%  $< -2$  LAZ. The majority of  
166 underweight measures came from the Pakistan peri-urban & village cohorts. On average  
167 children were already below expected length by age 1 month and dropped to an average  
168  $2SD$  below the mean by the second year of life, 14% were already  $< -2$  LAZ (Figure 1).  
169 Their weight showed a similar but less extreme pattern. Weight for length rose in the first  
170 few weeks and remained close to expected levels for the first 6 months, then dropped  
171 slightly below thereafter.

172 The associations of WAZ and WVZ with subsequent wasting and stunting were shown in  
173 Table 2. For both wasting and stunting, after adjustment for age, sex, country of origin and  
174 already being wasted or stunted, respectively, WVZ2 was a stronger predictor than WVZ6,  
175 but both were lower than for WAZ alone. The model including WVZ2 as well as WAZ had  
176 marginally higher prediction performance for subsequent wasting (AUC gain 1.5%) than  
177 the model with only WAZ, but for predicting stunting: the model including WAZ and  
178 WVZ2 the AUC was almost unchanged (AUC gain 0.4%).

179 The associations for those who were not wasted or stunted at the time of the attained WAZ  
180 (incident cases) and for children who were already stunted or wasted are shown in Table 3.  
181 This revealed that, although both WAZ and WVZ2 were more predictive of incident cases

182 than persistence in prevalent cases, WVZ essentially added no extra prediction of incident  
183 cases. It did, however, increase the prediction of recovery from wasting (change in AUC  
184 9.8%) but only marginally from stunting (change in AUC 1.4%).

185 Associations of growth parameters with mortality are shown in Table 4. Adjusted just for  
186 age, sex, and country of origin (Model 1), WAZ, WLZ, and WVZ (6-month interval) were  
187 separately associated with lower mortality risk, while WVZ (2-month interval) and LAZ  
188 were not significantly associated with mortality. Among combinations of the 3 significant  
189 predictors, the combination of WAZ and WVZ (6-month interval) yielded the strongest  
190 prediction (C-index 0.8842), slightly greater than using WAZ (C-index 0.8647) or WVZ-6  
191 alone (0.8718) but the overall gain was still only 2%.

192 Supplementary Tables 1-3 show the analysis of potential interaction with age group. On  
193 predicting wasting, in the combined model, WAZ showed a 12% increased association  
194 (Ratio of RR = 0.88; 95% CI 0.81-0.95) in older children, while WVZ-2 was 6% greater  
195 (Ratio of RR = 1.06; 95% CI 1.01-1.12). There was no evidence for age interactions for  
196 predicting stunting and mortality.

## 197 **Discussion**

198 This study used data from three cohort studies to describe the extent to which recent  
199 changes in weight (velocity) added to the accuracy of prediction of later wasting, stunting  
200 and mortality, compared to a single weight alone. Surprisingly, combining weight velocity  
201 with attained weight added little or no additional predictive value, except in predicting  
202 whether already wasted children would recover.

203 This finding is in keeping with the limited number of other papers that have directly  
204 compared velocity with single measures, mostly in relation to mortality. Two early studies  
205 found that attained weight and height for age performed better than weight or height  
206 velocity as discriminators of mortality (20, 21), while a more recent paper found little  
207 difference between various single measures and velocity (22). One group has found  
208 velocity to be a stronger predictor of mortality in one data set (23), but in a second set they  
209 found no increased prediction when adding velocity (24). In another study, the same group  
210 also considered predicting wasting and stunting and found that velocity increased  
211 prediction up to age 2 months (24), but not at later ages (9).

212 There are many possible reasons why velocity may not be a useful discriminator. Velocity  
213 is vulnerable to error as it requires two measurements, both with a degree of measurement  
214 error (e.g. instrument imprecision and differences in procedures) plus the variability of  
215 weight over short intervals due to ingestion and elimination (25). Although these data were  
216 collected as part of research studies, they were largely collected in real life situations where  
217 one could speculate that at one visit a young child might have just eaten and drunk largely,  
218 while at the next they might have fasted or just emptied their bowels or bladder. These  
219 combined are thus inevitably less accurate than a single weight measurement. Thus, a child  
220 could show a large meaningless variation ('noise'), from one time to the next that has no  
221 long-term significance, but has the potential to drown out true meaningful variation -  
222 the 'signal' (25).

223 It must also be born in mind that weight velocity is calculated between two measurements,  
224 and in utero growth could not be measured. If the decline in growth occurred in utero, the  
225 size at birth may be low, but there will be no observable decline in velocity. As a group,

226 these children had a high prevalence of stunting, illustrated by the fact that the mean LAZ  
227 in the 2nd year was close to -2 (i.e. 2 SD below mean). It has been well shown that a high  
228 proportion of stunting has already occurred very early in the first year (26). Other studies  
229 have shown that in uncompromised environments most children with slow intrauterine  
230 growth will show rapid catchup in the first 4 months (27), but in suboptimal circumstances  
231 recovery from stunting is rare (26). Thus most of these children remain at a similar centile,  
232 rather than a showing a detectable decline. This may also apply to the onset of wasting; a  
233 previous study in young infants that compared attained weight to velocity and skinfolds as  
234 a measure of low fat stores, found that a low weight was highly specific for both low  
235 velocity and low fat, but that only 40% of those with low fat had shown slow weight gain  
236 (28), suggesting that they had started life with low fat stores and failed to acquire fat, rather  
237 than faltering.

238 Finally, it should be recognised that velocity may vary rapidly over time, resulting in  
239 varying measurements dependent on when the measurement was taken and the duration  
240 where velocity is calculated. A recent study collected daily weights in a group of Gambian  
241 infants used a spline curve to “smooth out minor day-to-day fluctuations” and found that  
242 short episodes of weight loss were common (mean 18 days), but these were followed by a  
243 similar period (mean 17 days) of catchup (29). Depending on the times when weights were  
244 collected, the same child might thus be labelled as having very low or very high velocity  
245 (Figure 2). Over a longer period, the net effect may have been slower than expected weight  
246 gain, but it had two sources of errors (25). This deviation can equally be detected in most  
247 children simply using weight for age, as this describes how much the weight deviates from  
248 the average for that age, with only one error. So weight for age will always tend to be more

249 accurate than the net velocity in initially average children, who will always be in a  
250 majority.

### 251 **Strengths and limitations**

252 This study was able to combine data from three cohorts with comparable measurements,  
253 which enabled us to answer the research question with more events than analysing each  
254 cohort separately. Nonetheless, this study is not without limitations. Firstly, as a group,  
255 these children had very high rates of stunting, but our earlier description of this cohort  
256 showed that while incident wasting was higher than expected, it was much less common.  
257 (12). So in this setting, variations in WAZ may be more likely to reflect slowing of growth  
258 rather than short-term weight loss. Data collection was carried out more than 20 years ago  
259 when mortality and severe acute malnutrition were generally more common in the area.  
260 Thus, these findings may not be representative of modern populations. We have not  
261 considered the potentially modifying role of diet, in particular duration of breast feeding  
262 and age of first solids, environmental conditions or socioeconomic conditions, as these  
263 were not consistently collected between data sets. Similarly, we have not considered the  
264 role of birth length and weight in the analysis, as this study focus on the prediction value of  
265 weight velocity in practice, where birth length and weight is often not available.

### 266 **Implications**

267 The usual practice, which concentrates on the current weight, turns out to be the better than  
268 the proposed alternatives, as well as the easiest means of detecting wasting or predicting  
269 future adverse events. It is arguably reassuring that this analysis found no justification for  
270 incorporating weight velocity measures in routine screening programmes, which already

271 struggle to find time to process even single weights and heights. Incorporating the change  
272 from an earlier weight adds considerable complexity in practice. This would require the  
273 earlier weight to be accessed and the change interpreted, although recent developments in  
274 digital support for growth monitoring have the potential to make the interpretation easier  
275 (30).

276 The one possible exception for this is in the assessment of children with MAM or SAM (as  
277 indicated by wasting in this study), where the recent growth pattern substantially improved  
278 the prediction of persistence or recovery. This may lend support to the practice in many  
279 nutrition clinics of monitoring weight gain over time, not just attained weight, although this  
280 is not recommended by WHO (4).

281 In conclusion, although theoretically useful in detecting and predicting low growth and  
282 mortality, velocity measures add little or no predictive power, probably due to their  
283 increased imprecision compared to a single recent weight and the challenge of measuring  
284 them over the relevant time interval. The key anthropometric determinant of impending  
285 wasting, stunting, and mortality appears to be how far below the normal range the child's  
286 weight is, rather than how they reached that position.

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288 CMW, FPR, FKH designed the study and jointly-wrote the first draft. FPR and FKH  
289 analysed the data. RB, PA, SZ interpreted the data and critically revised the manuscript.  
290 CMW, FPR, RB, PA, SZ, and FKH have read and approved the final version of the  
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374 paperless age? The UK experience. *Archives of Disease in Childhood*. 2024;109(2):78-82.

375 Table 1. Baseline characteristics per attained weight WAZ categories

	Baseline WAZ categories		
	<-2	-2 to 2	>2
Total number of measurements, n (%)	4,167 (20.4)	15,953 (78.3)	265 (1.3)
Number (%) in each WAZ category by cohort:			
Malawi	842 (20.2)	2,734 (17.1)	4 (1.5)
South African and Pakistan middle class	154 (3.7)	3,968 (24.9)	201 (75.9)
Pakistan peri-urban & village	2,303 (55.3)	4,595 (28.8)	0
Pakistan urban slum	868 (20.8)	4,656 (29.2)	60 (22.6)

376 n: number; WAZ: weight-for-age z-score

377

378 Table 2. Associations of attained weight and weight velocity in one period, with wasting  
 379 and stunting in the next period.

Predicting wasting in next period	RR (95% CI)	AUC	Change in AUC	
Model 1 Wasting in this period	9.91 (8.3; 12.8)	0.872 (0.861-0.884)	REF	
Model 2 Wasting in this period WAZ	3.84 (3.1; 4.7) 0.66 (0.6; 0.7)	0.916 (0.908-0.924)	4.5%	REF
Model 3 Wasting in this period WVZ <sub>2</sub>	11.8 (10.5; 13.2) 0.76 (0.74; 0.79)	0.888 (0.878-0.899)	1.7%	
Model 4 Wasting in this period WVZ <sub>6</sub>	11.8 (10.5; 13.2) 0.83 (0.81; 0.86)	0.869 (0.858-0.880)	-0.3%	
Model 5 Wasting in this period WAZ WVZ <sub>2</sub>	4.06 (3.4; 4.9) 0.62 (0.58; 0.66) 0.74 (0.72; 0.76)	0.931 (0.927-0.941)	6.0%	1.5%
Predicting stunting in next period				
Model 1 Stunting in this period	6.61 (6.2; 7.0)	0.910 (0.905-0.915)	REF	
Model 2 Stunting in this period WAZ	5.30 (4.9; 5.7) 0.86 (0.85; 0.88)	0.929 (0.925-0.934)	2.0%	REF
Model 3 Stunting in this period WVZ <sub>2</sub>	6.62 (6.2; 7.1) 0.96 (0.95; 0.97)	0.913 (0.909-0.918)	0.3%	
Model 4 Stunting in this period WVZ <sub>6</sub>	6.63 (6.2; 7.1) 0.96 (0.95; 0.98)	0.912 (0.907-0.917)	0.2%	
Model 5 Stunting in this period WAZ WVZ <sub>2</sub>	5.22 (4.99; 5.6) 0.85 (0.84; 0.87) 0.95 (0.93; 0.96)	0.933 (0.929-0.937)	2.3%	0.4%

380 Data presented as risk ratios (RR) and their 95% CI using linear mixed-effect models. All  
 381 models were adjusted for age, sex and country of origin and then mutually adjusted for  
 382 all variables listed. WAZ: weight-for-age z-score; WVZ<sub>2</sub>: Weight velocity z-score 2-month  
 383 interval. WVZ<sub>6</sub>: Weight velocity z-score 6-month interval. All Relative risks were P<0.001

384 Table 3. Associations of attained weight and weight velocity with subsequent wasting and stunting in incident and prevalent existing  
 385 cases

	Risk of incident wasting in non-wasted children			Risk of continued wasting in already wasted children		
	RR (95% CI)	p-value	AUC	RR (95% CI)	p-value	AUC
Predicting wasting						
Model with WAZ only	0.49 (0.45-0.53)	<0.001	0.912 (0.902-0.922)	0.84 (0.80-0.88)	<0.001	0.707 (0.683-0.733)
Model WVZ2 only	0.70 (0.70-0.70)	<0.001	0.916 (0.604-0.928)	0.83 (0.79-0.86)	<0.001	0.743 (0.718-0.769)
Model with both			0.914 (0.902-0.925)			0.805 (0.783-0.828)
WAZ	0.46 (0.42-0.50)	<0.001		0.83 (0.79-0.87)	<0.001	
WVZ2	0.67 (0.64-0.70)	<0.001		0.82 (0.79-0.86)	<0.001	
Gain in AUC by adding WVZ to model			0.2%			9.8%
	Risk of incident stunting in non stunted children			Risk of continued stunting in already stunted children		
Model with WAZ only	0.52 (0.49-0.55)	<0.001	0.837 (0.826-0.848)	0.94 (0.92-0.96)	<0.001	0.761 (0.742-0.780)
Model WVZ2 only	0.89 (0.86-0.91)	<0.001	0.832 (0.820-0.844)	0.99 (0.97-1.00)	<0.001	0.678 (0.656-0.701)
Model with both			0.831 (0.825-0.847)			0.775 (0.757-0.793)
WAZ	0.49 (0.46-0.52)	<0.001		0.93 (0.91-0.96)	<0.001	
WVZ2	0.84 (0.81-0.91)	<0.001		0.98 (0.97-1.00)	0.02	
Gain in AUC by adding WVZ to model			-0.1%			1.4%

386  
 387 Data presented as risk ratios (RR) and their 95% CI using linear mixed-effect models. The created groups were: people who had  
 388 wasting/stunting at baseline (yes/no) vs those who developed later (yes/no). The main analyses were adjusted for age, sex, country of  
 389 origin and wasting/stunting status in the last period. WAZ: weight-for-age z-score; WVZ: Weight velocity z-score.

390 Table 4. Associations of attained height, weight and weight velocity with mortality

	HR (95% CI)	p-value	C-index (95% CI)	Change in C-index
Each variable added singly				
WAZ = Model 1	0.84 (0.73-0.98)	0.023	0.8647 (0.8263-0.9032)	REF
WLZ	0.83 (0.72-0.97)	0.02	0.8626 (0.8234-0.9018)	-0.2%
LAZ	0.91 (0.77-1.08)	0.30	0.8592 (0.8203-0.8980)	-0.6%
WVZ (2-month interval)	0.93 (0.80-1.06)	0.30	0.8606 (0.8226-0.8985)	-0.4%
WVZ (6-month interval)	0.78 (0.69-0.88)	<0.001	0.8718 (0.8325-0.9110)	0.7%
Model 2 (mutually adjusted)			0.8842 (0.8479-0.9205)	2.0%
WAZ	0.83 (0.72-0.96)	<0.001		
WVZ (6-month interval)	0.78 (0.69-0.88)	<0.001		
Model 3 (mutually adjusted)			0.8794 (0.8426-0.9162)	1.5%
WLZ	0.77 (0.66-0.89)	<0.001		
WVZ (6-month interval)	0.75 (0.67-0.85)	<0.001		
Model 4 (mutually adjusted)			0.8655 (0.8270-0.9041)	0.1%
WAZ	0.92 (0.74-1.13)	0.40		
WLZ	0.89 (0.72-1.10)	<0.001		
Model 5 (mutually adjusted)			0.8801 (0.8435-0.9167)	1.5%
WAZ	0.98 (0.80-1.21)	0.90		
WLZ	0.78 (0.63-0.96)	0.02		
WVZ (6-month interval)	0.75 (0.67-0.85)	<0.001		

391

392 Data presented as hazard ratios (HR) with their respective 95% CI using Cox proportional models. All analyses were adjusted for age,  
393 sex, and country of origin. Total measurements: 20,385 and deaths: 98. The c-index is an indicator of risk prediction which estimates  
394 the probability of concordance between observed and predicted responses. Values close to 0.5 are equivalent to a random guess while  
395 1.0 is equivalent to perfect prediction. WAZ: weight-for-age z-score; WVZ: Weight velocity z-score; LAZ: length-for-age z-score;  
396 BMI: body mass index.

397 **Figure Legends**

398

399 Figure 1. Age trend of growth indicators by sex.

400

401

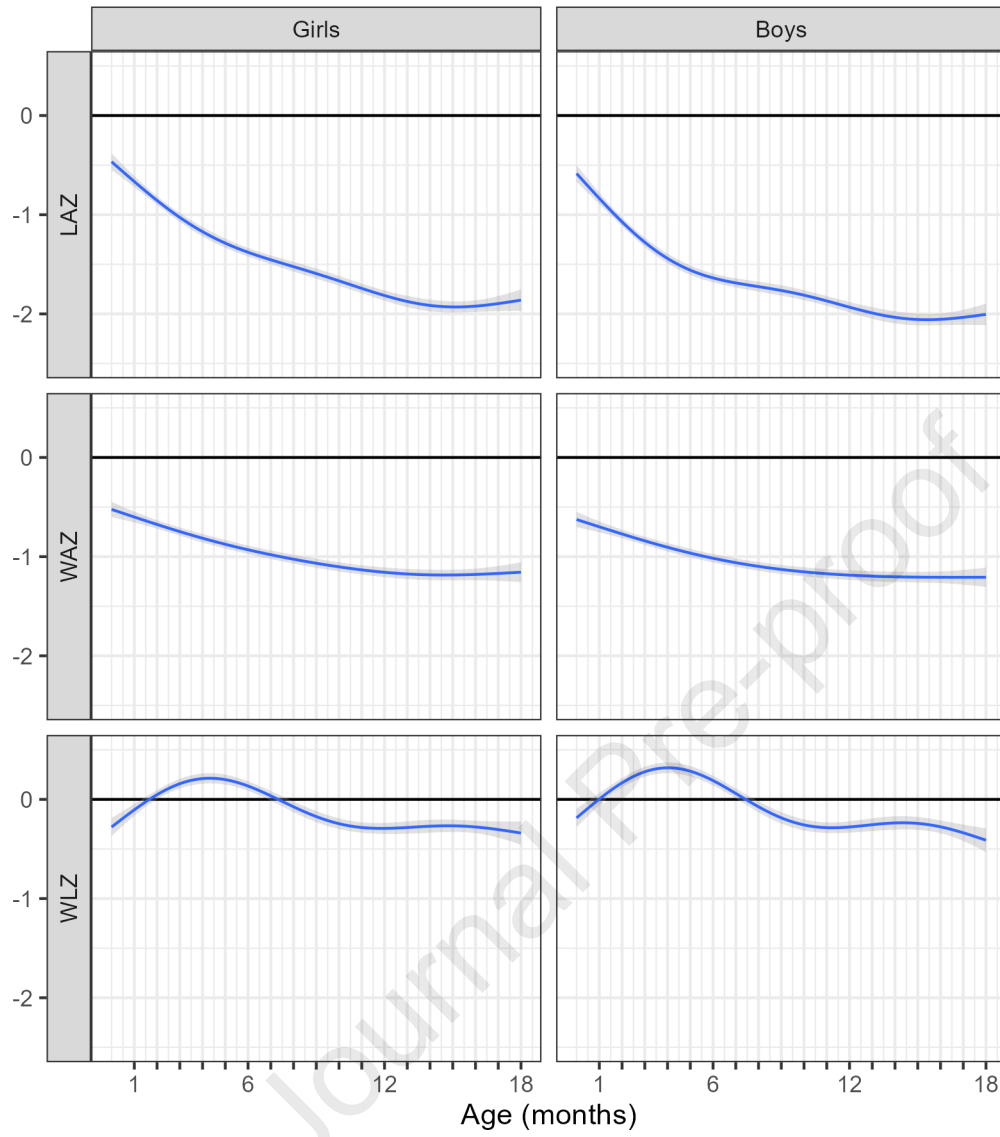
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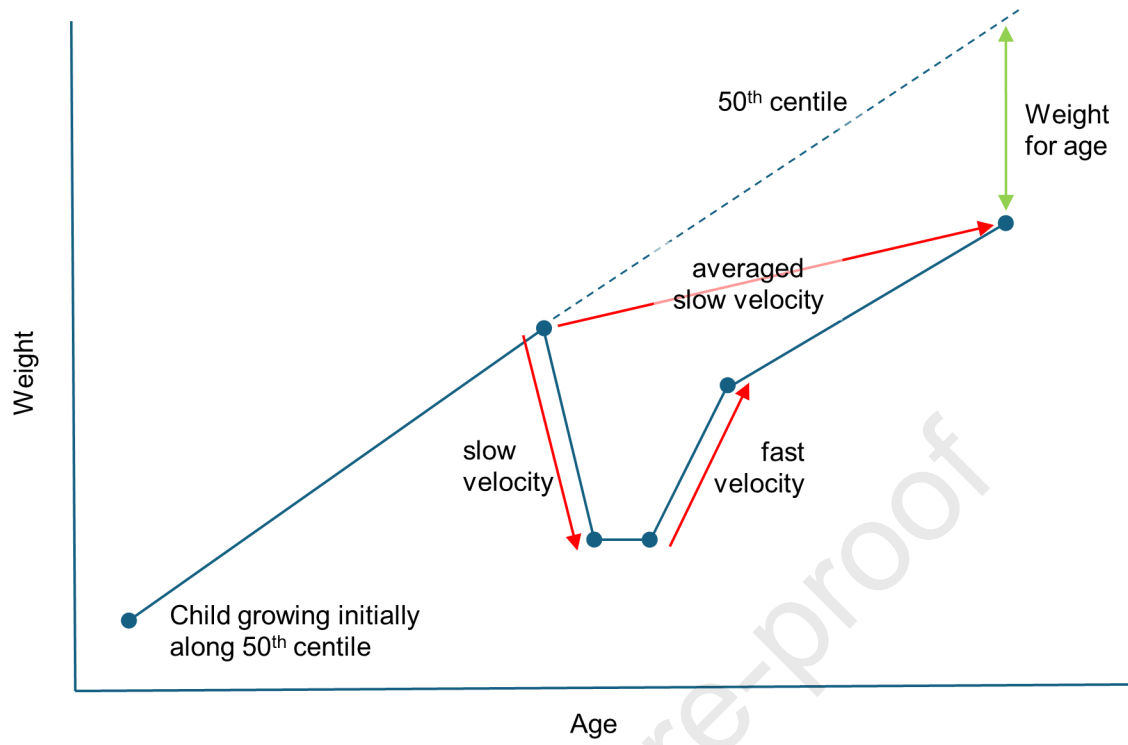
403 Figure 2: The challenge of assessing weight gain across an episode of weight faltering

404 A child showing slow, then fast velocity, averaged to a relatively slow velocity over a  
405 longer time, but with two sources of error. The last weight-for-age z score describes how  
406 much the weight deviates from the average for that age, with only one source of error

407







**Declaration of interests**

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

The authors declare the following financial interests/personal relationships which may be considered as potential competing interests:

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