



Response to Cummins and Finaret (2019)

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2 Response to Cummins and Finaret (2019)
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27 Dear Dr. Turner, Editor-in-Chief of the American Journal of Physical Anthropology,
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29 We thank Joseph Cummins and Amelia Finaret for their interest in our article (Comandini et al.,
30 2019) and their insightful comments, which allowed us to further discuss the issue of age
31 imprecision in nutritional studies.
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34 In this response, we aim to stress some points that interfere with the analysis of the three major
35 concerns highlighted by the authors.
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38 **1) Nature of Measurement Error** 39

40 We agree with J. Cummins and A. Finaret when declaring that in our research (Comandini et al.,
41 2019), we did not “mimic the methods used by large international demographic surveys.” This
42 would have been almost impossible, as our study was not based on households sampling as in the
43 surveys, but was realized within institutions where many children were boarding, some of them
44 from many years, many were orphans, some came from the main slum of Kampala, and had been
45 selected there by the school for being the more in need (i.e., in most cases, without a protective
46 family). However, the longitudinal approach of the research allowed us to collect repeated
47 information on child age from school-registers and social workers archives, and to integrate them
48 with data obtained by interviewing the teachers, the nurses, and in some cases the mothers.
49 Actually, we aimed to consider statistically the magnitude and impact of age imprecision in a large
50 sample of school-children, which is an interesting but poorly studied group of children.
51 In fact, the Demographic and Health Surveys (DHS), as well as the Multiple Indicators Clusters
52 Surveys (MICS), focus only on adults and children under the age of five. Anyway, although we
53 agree on the excellent quality of data from DHS and MICS surveys, they meet age errors too, as
54 indicated by the frequent age heaping (Larsen et al., 2019), and by cases exclusion due to invalid
55 age data. Indeed, non-response rates for child ages as high as 30% have been detected in DHS
56 surveys (Larsen et al., 2019). The problem is particularly relevant in countries where the practice of
57 registering children at birth is lower, such as Uganda. As an extreme case, the MICS reports on the
58 Northeast Zone of Somalia, a region with the lowest rates of birth registration in the world, omitted
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2 all the information on child undernutrition because it was “extremely difficult to get correct age-
3 related data” (UNICEF Somalia and Ministry of Planning and International Cooperation, 2014).
4 Furthermore, noteworthy, the values of age data omission declared by the DHS and MICS are likely
5 underestimated, as the standard surveys refer to the population living in households, while people
6 living in the slums are generally under-represented (Elsey et al., 2018).

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8 The systematic and random error in age detected in nutritional studies has a significant impact on
9 malnutrition estimates. We thank J. Cummins and A. Finaret for their clear synthesis on the effects
10 of age imprecision, particularly of random error in the month of birth and of age heaping. We also
11 thank them for noticing that Larsen et al. (2019) detected an impact of age imprecision on stunting
12 similar to that estimated in our sample, which is about one percentage point. This degree of effect
13 magnitude is lower than the one calculated in previously published studies, using simulations of
14 DHS data on children under five (Comandini et al., 2016; 2017). Indeed, this is not surprising
15 because the effect of age imprecision depends on several factors, such as the extent and spread of
16 the error within the sample, and the child’s age, being lower in older children (Comandini et al.,
17 2016; 2017).
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20 **2) Study Population**

21 We agree with J. Cummins and A. Finaret that school-children generally show higher stunting
22 prevalence than younger ones. However, the direct comparison of stunting prevalence between our
23 sample (11.9–12.7%) and the Ugandan population of children under five (28.9%) surveyed by
24 UDHS (UBOS & ICF, 2018) is not appropriate, as we never declared or intended to imply that our
25 results were representative of children at risk for malnutrition in the whole Uganda, nor in the
26 Ugandan regions where the sampling has been done. Also, the employed bootstrap method
27 does not allow generalization to the population level. Indeed, according to the 2016UDHS (UBOS &
28 ICF, 2018), the distribution of malnutrition in Uganda is strongly variable, with some areas, such as
29 the Karamoja region, showing levels of malnutrition twice those detected in the Kampala region
30 (35.2 vs 18.1%), that is the more common provenance of the children in our sample. Furthermore,
31 though the majority of children in our sample came from impoverished families, they have attended
32 private schools thanks to Bhalobasa Association help. In these schools, children can benefit from
33 hygienic and dietary support, according to practices studied and improved during the research, and
34 briefly described in our article.
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37 That said, we are aware that children with a better nutritional status are less affected by age mis-
38 measurement than more malnourished children, as their z-scores are higher and less interested by
39 random oscillations around the - 2SD cut off threshold for undernutrition. Indeed, along with the
40 influence mentioned above of the different magnitude of imprecision and the different children
41 ages, mean nutritional status is another relevant factor in the variable effect of age misreporting on
42 nutritional status estimates.
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45 **3) Levels v. Differences**

46 We agree entirely with J. Cummins and A. Finaret on the significant interference of age error with
47 understanding the etiology of malnutrition, especially of stunting, and the health impact of either
48 policies, programs, or interventions.

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50 Until the goal of registering all children at birth will be satisfactory approached, a more solid
51 consciousness of the relevance of age data within the nutritional and anthropological research
52 community is to be hoped. A careful procedure to collect age data is strongly advisable. Further,
53 reports on nutritional status in children and adolescents from low-income countries should report a
54 comment on the quality of age data. Such an approach would be particularly useful in studies on
55 nutritional status in school-children, where more scattered information is available. Indeed, the
56 majority of the existing knowledge originates from data collected using methodologies different
57 from the DHS ones, and without discussing the problem of age bias or imprecision. On the other
58 hand, regrettably, DHS and MICS data are not suitable to analyze nutritional status in school-
59 children, as these children show different growth patterns and are influenced by different causal
60 factors than those under-five.

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3 The authors had no conflicts of interest to declare.
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