Season of data collection of child dietary diversity indicators may affect conclusions about longer term trends in Peru, Senegal and Nepal

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Disclaimers: None

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Conflict of Interest Statement: Andrew L Thorne-Lyman, Leah E M Bevis, Helen Kuo, Swetha Manohar, Binod Shrestha, Angela KC, Rolf D Klemm, Rebecca A Heidkamp have no conflicts of interest.

Sources of Support: Support for this research was provided by the Bill & Melinda Gates Foundation, Seattle, WA and the Feed the Future Innovation Lab for Nutrition, which is funded by the United States Agency for International Development under grant number AID-OAA-L-1-00006. These funders did not play a role in design, implementation, analysis or interpretation of the data.

Abbreviations list: DHS, Demographic Health Surveys; FGS, Food Group Score; LMIC, Low and Middle Income Country; MAD, Minimum Acceptable Diet; MDD, Minimum Dietary Diversity for Children’ PoSHAN, Policy and Science for Health, Agriculture and Nutrition;; SD, Standard Deviation; WHO, World Health Organization

Supplementary material is available from the ‘Online Supplementary Material’ attachment.
Abstract

Background: The WHO-UNICEF minimum dietary diversity (MDD) indicator for children aged 6–23 months is a global monitoring indicator used to track multi-year population-level changes in dietary quality, but the influence of seasonality on MDD estimates remains unclear.

Objective: Examine how seasonality of data collection may influence population-level MDD estimates and inferences about MDD change over multiple survey years.

Methods: Three countries were selected with three or more consecutive years of MDD data collection including continuous national Demographic Health Surveys in Senegal (2012–2017, N = 12,183) and Peru (2005–16, N = 35,272) and the PoSHAN sentinel site seasonal surveys (covering 3 seasons/year) in Nepal (2013–2016, N = 1,309). MDD prevalence (≥5 of 8 food groups) and an 8-item continuous food group score (FGS) and 95% confidence intervals were estimated by month and compared for lean and non-lean seasons using Ordinary Least Squares (OLS) regression with dummy variables for year.

Results: The national prevalence of MDD was higher in Peru (75.4%) than in Nepal (39.1%) or in Senegal (15.7%). Children in Peru were 1.8% [coefficient: -0.0179 (-0.033, -0.002)] less likely to achieve MDD during the lean season. Similar seasonal magnitudes were observed in Senegal [-0.0347, (-0.058, -0.011)] and for Nepal [-0.0133 (-0.107, 0.081)]. FGS was about 0.1 item lower during the lean season in all three countries. In comparison, MDD increased by an average rate of only 4.2 and 4.4 percentage points per five years in Peru and Senegal respectively. Intake of specific food groups was stable across months in all countries with the provitamin-A rich food group exhibiting the most seasonality.

Conclusions: The magnitude of seasonal variation in MDD prevalence was smaller than expected but large relative to longer term changes. If large-scale surveys are not conducted in the same season, biased conclusions about trends are possible.

Key words: Dietary diversity, seasonality, season, Nepal, Peru, Senegal, child diets, dietary quality, indicator

Teaser text: Seasonality of the minimum dietary diversity indicator for children was explored Peru, Senegal and Nepal. If surveys are not conducted in the same season, biased conclusions about trends may be reached.
Introduction

Policies and interventions focused on improving household diet quality have become a major part of multisectoral nutrition programs in many countries (1). Accordingly, tracking indicators of diet quality for key populations is a growing priority among governments and development partners. The increase in usage of these indicators across sectors is striking: a 2010 review of agriculture projects identified no projects tracking these indicators, but by 2016, the use of diet or food consumption indicators had grown to 93% of 73 surveyed agriculture-to-nutrition projects (2, 3). Simple dietary diversity scores, which tabulate the total number of food groups consumed in a defined recall period (e.g. 24 hours), are of particular interest because they are relatively easy to collect and analyze compared to other diet quality measures and in near term may also be more responsive to programs than anthropometric indicators. Two of the most commonly used scores (Minimum Dietary Diversity for Child, MDD, and Minimum Dietary Diversity – Women, MDD-W) have been validated against nutrient adequacy in several contexts (4-8).

Minimum Dietary Diversity for children MDD is a core indicator in the WHO Global Nutrition Monitoring framework and national estimates are tracked over time for all WHO member states (9). Many LMIC have set national and subnational targets for MDD related to reducing the proportion of children with low MDD or Minimum Acceptable Diet (MAD), an indicator derived in part from MDD (10). Country profiles for the Global Nutrition Report, a global nutrition accountability report, include both MDD and MAD (11). Program evaluation guidance for USAID-funded programs also recommends MAD as an outcome indicator (12).

The global MDD indicator was first released by WHO-UNICEF in 2008 as part of a set of standard indicators to assess infant and young child feeding (IYCF) practices among children 0-23 months of age (2, 3). The original MDD indicator was tabulated using seven food groups including (1) grains, roots and tubers (2) legumes and nuts (3) dairy products (4) flesh foods (5) eggs (6) vitamin A rich fruits and vegetables and (7) other fruits and vegetables (13). In 2017, MDD was revised to include breastmilk as an eighth food group and
is now defined as the proportion of infants and young children consuming least 5 of 8 food groups in the past 24 hours (9).

One of the challenges inherent in the use of simple dietary indicators to track population-level improvements in diet quality across time is that seasonal patterns often influence food availability, affordability and accessibility in LMIC. Many countries experience seasonal lean seasons prior to the harvest in which lower availability of food and money adversely affects access to more diverse and costly foods. Prices of nutrient-dense foods may also vary by season (14). Infants and young children may also experience seasonal patterns of illnesses such as malaria, respiratory infections and diarrhea that adversely affect their appetite and feeding practices by parents (15).

Ideally, surveys used to generate national and subnational MDD estimates would be conducted at the same time each year to account for seasonality. However, numerous factors can prevent this from happening such as political events and resource mobilization constraints. High-level monitoring of MDD across time in LMIC frequently relies on information collected through large-scale multitopic household surveys. Necessary questions have been included in Demographic and Health Surveys (DHS) core questionnaire since round 6 (2009) and are included in the UNICEF Multiple Indicator Cluster Survey core questionnaire. Multiple countries in sub-Saharan Africa also implement annual national and subnational nutrition SMART surveys that measure MDD. Large surveys are frequently conducted over several months which may span different seasons, with possibilities of variation in the timing of data collection by survey wave, resulting in estimated annual trends that are potentially influenced by seasonal patterns.

However, few analyses have examined the influence of seasonality on high level estimates of dietary diversity indicators and none to our knowledge have explored the possible implications of the timing of data collection for global and national monitoring of progress on MDD over multi-year periods. In fact, a recent review of the seasonality of different causes of malnutrition in South Asia identified only one study examining seasonality of child diets, a study that pre-dated the development of the MDD indicator (16).
Using multi-year datasets collected across seasons in three countries with different child diets and seasonal patterns (Nepal, Peru and Senegal) our study aims were to:

1. Examine the degree to which variability in the season of data collection may influence the national prevalence of MDD across survey years in three countries.
2. Understand the relative importance of seasonal vs. annual changes in this indicator.
3. Explore interactions between seasonality of dietary diversity and rural/urban residence.

Methods

Description of data sources

Three different data sources were used for these analyses. Continuous DHS datasets were used for Senegal and Peru. For Nepal, data was used from a set of seasonal sentinel site surveys administered as part of the Policy and Science for Health, Agriculture, and Nutrition (PoSHAN) community studies. Across the three countries, we retained households with children 6-23 months of age having the available dietary data needed to construct the MDD indicator.

Peru

The continuous DHS in Peru was the first DHS to collect data at multiple timepoints in a year and across consecutive years and its methods are described in detail elsewhere (17). This analysis uses data from January 2005 to December 2016, available from the DHS Program and Peru National Institute of Statistics websites (18,19). In brief, the survey used a two-stage household stratified cluster sampling method and employed a cross-sectional design. The sample was stratified by urban/rural areas within 25 regions (Peru’s 24 departments and one other province) and conducted over a ten-month period. The Peru continuous DHS was nationally representative starting in 2004 and also representative at the department level from 2008 onwards. Our analysis accounted for the two-stage stratified sampling procedure and changes in the sampling design (i.e. changes in
survey strata) across years by adjusting for year-specific, stratified child weights. Sample size (number of children 6-23 months of age) is listed by month and year in **Supplementary Table 1**.

**Senegal**

Data from the continuous DHS in Senegal from 2012-2017 were accessed from the DHS Program website (18). The DHS survey used a two-stage household stratified cluster sampling method and also employed a cross-sectional design. The sample was nationally representative for all years and stratified by urban/rural areas within the 14 regions of Senegal. Detailed descriptions of the methods used are reported elsewhere (20,21). However, as with the continuous DHS survey in Peru, the survey strata changed partway through the continuous survey in Senegal, in 2013. Our analysis again accounts for both the two-stage stratified sampling procedures and the changes in strata by adjusting for year-specific, stratified child weights. Sample size ranged from 772 in 2004 to 6,404 in 2016, with a total of 12,183 children over the surveyed period. Sample size is presented by month and year in **Supplementary Table 2**. Data from 2012 and 2013 were combined into a single year (labeled 2013) in all analysis, as both represented half a year of continuous data from different seasons.

**Nepal**

The Nepal dataset included data on 1,309 unique dietary assessments from 615 children 6-23 months of age. The methods for collecting the seasonal data from PoSHAN Community Studies of Nepal are described in detail elsewhere (22–24). In brief, the larger PoSHAN study involved the collection of national data from 2013-2016, using a two-stage systematic random sampling of seven village development committees (VDC’s) in each of Nepal’s three agro-ecological zones: the Mountains, Hills, and Tarai (plains). Subsequently, three wards (the smallest administrative unit in Nepal) were randomly chosen per VDC (22). Within each of these wards, all consenting households with children <60 months of age and all consenting households with recently married couples without children were surveyed. Within this larger sampling frame of 21 VDCs and 63 wards, one VDC was selected in each zone as a “sentinel site.” These 3 sentinel sites were chosen based on their representativeness of the zone according to multiple indicators from the Central Bureau of Statistics (22).
Starting in 2013, households in the sentinel sites were prospectively followed across seasons for two continuous years, with data collected three times per year: in the monsoon lean season during which rice and most other crops are grown (June/July/Aug), in the post-harvest season (Late Aug-Sep) and in the winter lean season (January/February). These three seasonal categories are consistent with those typically used in the literature for South Asia (16,25). Additionally, data were collected during the 2016 monsoon season. Thus, the survey covers 7 time periods: 2 post-monsoon seasons (2013, 2014) and 2 winter seasons (2014, 2015) and 3 monsoon seasons (2014, 2015, 2016), and both cross-sectional and longitudinal analyses are possible as households and children were tracked over time using longitudinal identifiers. New children could enter the seasonal sample surveys only during the monsoon seasons in 2014 and 2016. The result is that the average child was seen 2.1 times. Forty-five percent of sample children are viewed only once, and 21%, 13%, 21% and 1% are viewed twice, three times, four times, and five times, respectively. Similarly given that we restricted the analysis to children 6-23 months, many children enrolled in 2013 aged out by 2014, and all of them aged out by 2016. Sample sizes for each period are presented in Supplementary Table 3.

Analysis

Outcome definition:

Minimum dietary diversity (MDD) was calculated for children 6-23 months of age using standard classifications generated from the report “Indicators for assessing infant and young child feeding practices: Definitions and Measurement methods” produced by the World Health Organization and UNICEF (26). Food groups included breastmilk, grains/roots/tubers, legumes/nuts, dairy products, flesh foods, eggs, provitamin A rich fruits and vegetables and other fruits and vegetables. A child was defined as achieving MDD if they consumed at least 5 of the 8 food groups in the past 24 hours resulting in a binary outcome variable. We also generated a continuous Food Group Score (FGS) variable as a count variable ranging from 0-8 of the total number of food groups the child consumed including breastmilk.

For Nepal, a 24-hour food frequency questionnaire was administered to the primary child caretaker to capture child consumption of 31 commonly consumed food items. These were then consolidated into the food groups.
groups noted above except for breastmilk. A 7-day recall was administered to assess the number of times breastmilk was consumed; children with any consumption were counted as “yes” and contributed to the dietary diversity score, those with no consumption over 7 days were counted as “no”. In Peru and Senegal, the continuous DHS survey contained questions about the consumption of 25 and 17 different food groups, respectively, in the last 24 hours. These were designed to capture the majority of foods consumed by children in each country. The DHS surveys also indicated whether the child in question was currently breastfeeding. The information on food groups and breastfeeding was then consolidated into the same 8 food groups mentioned above to create FGS and MDD.

Definition of seasons:

In Nepal, we compared prevalence of MDD and mean FGS across the three seasons defined by the original PoSHAN sampling design, described above. We hypothesized that both would be lower in the monsoon lean and/or in the winter lean season, as compared to the postharvest season.

For Senegal and Peru, we used two different approaches to defining season. The first was informed by our primary goal of understanding how the timing of data collection within a calendar year might influence estimates of MDD prevalence and mean FGS. To that end, we defined “lean season” as the three sequential months of lowest dietary diversity, recognizing that large national surveys often take multiple months to complete. We then compared MDD prevalence and mean FGS across these three months (lean season) vs. the rest of the year (non-lean season). In Peru, the lean season corresponded to May-July and in Senegal April-June. For Senegal, our definition of lean season differed from the conventional definition that relates to food availability (June-September), because this period was not found to be the period of lowest dietary diversity.

A secondary seasonal contrast for these two countries emerged from a hypothesis that vegetables and fruits may be less available during the dry season, potentially leading to lower dietary diversity during that season. This hypothesis was tested by comparing MDD and FGS by “rainy season” vs. “dry season”. We defined the rainy season as October-April in Peru and July-September in Senegal based on maps of rainfall patterns by month presented in Supplementary Figures 1-3, created in R using the Era-Interim data from the European Center for Medium-Range Weather Forecasting, during the period 1978-2018.
Analysis

All estimates for Peru and Senegal were weighted averages or weighted regressions, according to the two-stage sampling scheme. Adjusting for changes in sampling stratification across pooled cross-sectional data is necessary for valid inference. Our adjustment follows DHS protocols for pooled cross-sectional analysis, adjusting for changes in both survey strata and probability sampling units (“clusters”) over the years. The resulting survey weights reflect probability of being sampled in any given year (this probability naturally falls with population size), allowing averages or statistics to retain their nationally representative status. While dietary diversity variables are known to vary by wealth and child age, we did not include such variables in our models because the explicit goals of our analyses were to investigate whether the prevalence of MDD varies by season to a degree that it could drive spurious variation across survey rounds over time. The 95% confidence intervals for all plots reflect standard errors clustered at the DHS cluster level (Peru and Senegal) or ward level (Nepal), and regression standard errors are clustered at these same levels.

First, the prevalence of MDD and mean FGS with 95% confidence intervals were calculated, plotted and visually inspected by month, by year, and for each season for all three countries. Next, we examined seasonality in dietary diversity indicators using ordinary least squares regression. In all regressions, dummy variables for years were included to parse the effect of season vs. more general temporal change. Analyses for Peru and Senegal were weighted according to the two-stage sampling scheme with errors clustered at the DHS cluster level. Simple OLS analyses for Nepal had errors clustered at the ward level. Additional OLS regressions for Nepal included child fixed effects (random intercepts) with errors clustered at the child level. This fixed effect specification further isolates the role of seasonal variation by controlling for any unobserved, time-invariant child characteristics that influence dietary diversity; effectively, children are being used as their own “controls” when observed both during and outside of the season of interest. Student t-tests were used to test for seasonal differences in outcomes in all countries and an F-test was used to test the joint influence of the three agricultural seasons in Nepal. In a last specification, we also examined an interaction between seasons and urban/rural residence, to explore whether seasonal patterns could differ across these strata, which represent very different
food environments. For all regressions, tests of statistical significance are denoted at the 1%, 5%, and 10% level. Data was analyzed in SAS 9.4, STATA 14.2, and Rstudio 1.1383 running R 3.4.3.

Consent and IRB approvals:

Ethical approval for the PoSHAN study in Nepal was granted by the institutional review board at the Johns Hopkins Bloomberg School of Public Health (JHBSPH) and from the Nepal Health Research Council. Ethical oversight of the Demographic and Health Surveys for Peru and Senegal is managed by the IRB at ICF International, and use of the datasets was deemed not human subjects research by the IRB at JHBSPH as these are de-identified datasets.

Results

The sample included 35,272 children aged 6-23 months from Peru, 12,183 from Senegal, and 615 children representing 1,309 observations over time from Nepal. The temporal distribution of children with MDD/FGS measures is presented in Supplementary Tables 1-3. These tables show the presence of certain data gaps for each of the datasets that were due to the design and implementation of multiple data collection rounds: for example, in Peru, the period from 2004 to 2008 had fewer observations for the months of October to December while 2007 to 2016 had few observations from January and February. Similar gaps exist in the data for both Senegal and Nepal.

Selected characteristics of the three country samples are presented in Table 1. About 1/3 of the Peru sample was collected from rural areas, compared with about 2/3 of the Senegal sample. The Nepal sample was made up mainly of rural areas although one ward was semi-urban. Peru had the highest prevalence of MDD in the full sample, at nearly 65% compared with about 10% in Senegal and 39% in Nepal.

Figure 1 presents annual trends in estimates of MDD prevalence and mean FGS for Peru and Senegal at the national level. In Peru, annual prevalence of MDD rose steadily between 2005 and 2013 where it then plateaued; the highest estimate of MDD prevalence was observed in 2014 at 77.9% (95% CI 76.0, 79.9). The mean FGS for Peru exhibited a similar rise through 2013, peaking in 2015 at 5.45 food units (95% CI 5.39, 5.52). For Senegal, the annual prevalence of MDD stayed constant at ~15% during 2013-2015, fell to just under
12% in 2016, then rose in 2017 to 18.8% (95% CI 16.8, 20.7). Interestingly, annual estimates of mean FGS showed a fairly steady decline between 2013 and 2016 in Senegal, despite the fact that MDD stayed constant for the first three years. Then, like MDD, mean FGS rose in 2017 to 18.8% (95% CI 16.8, 20.7), almost exactly where it had started in 2013. Therefore, while a clear overall improvement was observed in dietary diversity in Peru, only minor year-to-year variation in dietary diversity was observed in Senegal, with very little (MDD) or no (FGS) overall improvement.

Figure 2 presents the prevalence of MDD and mean FGS disaggregated by month, pooled across all years/data collection points for all three countries. The prevalence of MDD exhibited monthly variation in all countries. For Peru, a slight seasonal dip was observed from May to July, where MDD prevalence hovered around 74%, followed by a rise in August. While a dip in MDD prevalence was also observed in January and February, very few observations exist in those months and so the confidence intervals around those estimates are too wide to draw inference about the true seasonal mean. Seasonality was much more pronounced in Senegal, where MDD prevalence estimates hovered between 15-20% in all months except April-June, where they dropped below 15%. MDD prevalence was lowest in June, at only 9.34% (95% CI 6.2, 12.6). In Nepal, the estimated MDD prevalence was bound within an approximate 12% point range across months, ranging from 34.7% (95% CI 11.0, 58.3) in May to 46.1% (95% CI 21.3, 70.8) in July. However, some months did not have data collection activity, and so it is possible that this range of 12% could be larger if the data collection period had spanned all months of the year. Confidence intervals were also widest in Nepal, due to the relatively smaller sample size of this survey.

In Peru and Senegal the national prevalence of MDD was 1.9 and 4.1% lower in the lean season than the rest of the year, respectively. In both countries, the mean FGS was about 0.1 units lower in the lean season than during the rest of the year. (Figure 3). In Nepal, the post-harvest season had a lower seasonal estimate than the other two seasonal data points, although all estimates were bound by wide and overlapping 95% confidence intervals. For both Peru and Senegal, MDD prevalence and mean FGS were similar for the rainy and non-rainy seasons (Supplementary Figure 4).
Using regression analysis to separate seasonal variation from year-by-year variation, we continued to find evidence of small but consistent seasonal patterns suggesting that dietary diversity was lower in the lean season. Analyses of national datasets from Peru and Senegal suggested child MDD prevalence was 1.8 and 3.5% lower during the lean season, respectively (Columns 1 and 3, Table 2). Simple OLS analysis suggests that in Nepal, child MDD was similarly 1.3% lower during lean season (Column 5, Table 2) and child fixed effects analysis showed similar results (Supplementary Table 5). However, confidence intervals were much wider in Nepal due to the smaller sample size (11% and 4% of the Peru and Senegal sample sizes, respectively), and thus coefficients were not statistically significant. For FGS, in all three countries, children consumed about a tenth of a food group less during the lean season, although findings were statistically significant (p<0.01) only in Peru and in Senegal (Table 2). No significant differences were observed in Peru or Senegal for rainy vs. dry season (Supplementary Table 4).

In regression models testing for the interaction of urban/rural residence and lean season, significant seasonal patterns in both MDD and FGS were apparent in the rural areas of Senegal and Peru, but not in urban areas (Table 3). In both countries, prevalence of MDD fell by 3% during the lean season in rural areas. Both prevalence of MDD and mean FGS were also higher in urban households, who consumed half a food group more than rural households, on average. Both prevalence of MDD and mean FGS were marginally higher during the rainy season in rural areas (Supplementary Table 6), even though little difference was observed by rainy season at the national level or in urban areas. We also explored the possibility that differences in the timing of visits by survey teams to different regions in Peru and Senegal might be responsible for observed seasonality by adding regional fixed effects to these rural/urban regressions (Supplementary Table 7). Results did not meaningfully differ from those in Table 3 suggesting that regional differences in survey implementation timing are not responsible for seasonality.

The proportion of children consuming individual food groups was also explored in each country (Figure 5). Results from Peru suggested that the consumption of most food groups was stable around the year. In Senegal, breastmilk consumption was negatively correlated with staple consumption. Additionally, we see that while consumption of both flesh foods and provitamin A rich foods dropped during August-October,
consumption of legumes and nuts rose during this same period, suggesting a buffering effect. These trends were even stronger in rural areas. In Nepal, the rise in FGS in July appeared driven by a rise in provitamin A fruits and vegetables and eggs, though this rise was accompanied by a decrease in dairy consumption in July. In Senegal, a similar inverse relationship was observed between breastmilk consumption and staple consumption as was observed in Peru.

Discussion

This study helps to fill an important gap in the measurement agenda related to the use of the MDD indicator for high-level monitoring of the quality of diets of infants and young children in LMIC. Much of the validation work related to the child MDD indicator to date has focused on its ability to serve as a proxy for nutrient adequacy (27,28). Subsequently, refinement of the indicator from a seven-item to an eight-item score including breastmilk took place to improve its utility as a more comprehensive indicator of children’s diets during the 6-23 months period (10,11). However, even though patterns in food insecurity and food availability are known to be seasonal in many rural contexts, we believe that this is the first study to examine the seasonality of the MDD indicator from the perspective of understanding its validity for monitoring high-level changes in diet quality over time (i.e. the type of monitoring done by UN agencies, governments and global accountability initiatives).

Despite the vastly different prevalence of MDD in the three contexts, remarkable similarities were observed in the magnitude of lean season effects on MDD. Although the seasonal patterns were strongest in Peru, the range of lean season coefficients for MDD across the three settings was fairly consistent, ranging from 2% to 4%. Even greater consistency was observed across all three countries in the magnitude of seasonal effects on the FGS indicator. In both Peru and Senegal, no significant seasonality in either MDD or FGS was present for children in urban areas.

Growing attention is being paid to country level progress related to dietary shifts over time, and child MDD represents a key indicator for many countries and agencies as it is easily collected and can be included in major large-scale surveys such as the DHS and MICS. These surveys are often spaced at 5-year intervals or
longer, with the change in prevalence of sequential surveys used to infer trends. Some countries, including multiple countries in the West Africa region, conduct annual nutrition surveys between DHS and MICS years and infer MDD trends over a shorter interval. Yet the timing of such surveys within a calendar year is not necessarily calibrated to occur within the same months or season, as many factors may influence the timing of a survey going to the field.

Our research aimed to examine whether seasonal differences in the timing of surveys might exert a spurious influence on inference about national-level changes in MDD over longer periods of time, particularly if two surveys were not conducted in the same season. Often the period between national surveys is five years. Extrapolating the observed annual rates of change for MDD in Peru and Senegal per five years (by subtracting the last annual estimate from the first, dividing by the observation period and multiplying by five) we observed an estimated MDD increase in Peru by an average of 4.2 percentage points for Peru and 4.4 percentage points for Senegal per five years. These multi-year changes were not markedly larger than the average seasonal changes in MDD observed in these countries, which as noted above ranged from 2-4 percentage points. This suggests that the seasonal variability in MDD prevalence could indeed be large enough to introduce bias into long-term MDD trends, and lead to the wrong conclusions. The same was true for FGS; the average five-year rate of change in FGS was 1.15 food items in Peru (an improvement), and 0.02 food items in Senegal (a slight decline). Yet FGS varied across lean vs. non-lean season by 0.09 items in Peru and 0.14 items in Senegal, again suggesting that seasonality may introduce spurious influence on multi-year estimates of trends in dietary diversity, particularly in countries like Senegal experiencing slow change.

However, governments and large-scale projects often assume that much greater improvements in dietary diversity will be made over time than those we observed in the present study. For example, the nutrition-sensitive agriculture strategy in Ethiopia set an ambitious target for change in the MDD indicator from 9.8% to 40% from 2016 to 2020, with planned annual surveys to detect a change of approximately 6% per year.(29) The seasonal fluctuations in data collection across years observed in our study would exert only a minor influence on long-run prevalence estimates in countries where such rapid rates of change are occurring, but might still be important to consider if annual measurements were being compared. For many countries,
particularly those with a higher baseline MDD prevalence, the expected change between years would be much less, and even small seasonal fluctuations of the magnitude we observed could influence inference about progress over time.

Our study also revealed some insights into the comparative responsiveness of MDD and FGS to seasonality. Dichotomizing an indicator (such as FGS into MDD) results in the loss of information which is particularly important when considering change in indicators over time (e.g. across rounds of national surveys) or space (e.g. across regions within a country). MDD will only reflect change in dietary diversity that occurs across the FGS>=5 threshold. So, if dietary diversity is improving amongst segments of the population with much lower or higher mean FGS, that change may not be captured by MDD. This helps to explain why the drop in mean FGS experienced in Senegal between 2013 and 2015 was not mirrored by an accompanying drop in MDD prevalence. Because the average FGS is low in Senegal (2.9 over the entire survey period), changes in FGS are not necessarily reflected in changes in MDD, because they happen below the threshold of FGS=5.

This study responds to calls for further validation of the seasonal dimensions of dietary diversity scores, and is the first to our knowledge to explore the seasonality of the MDD indicator among children 6-23 months of age. Surprisingly few studies have examined the seasonality of diets of infants and young children in global settings. In South Asia, a recent comprehensive review of seasonal determinants of undernutrition in South Asia identified just one study from south Asia, from 1985, exploring seasonality of dietary intake among infants in Bangladesh, which pre-dated development of the MDD indicator.(15,16) A recent study conducted among 6-59 month old children and their mothers from Timor-Leste found evidence of lower dietary diversity during the dry and wet seasons compared with the dry and transition seasons, although a lower threshold of only 2 food groups was used to define low dietary diversity compared with the MDD indicator.(30)

However, a number of studies have been conducted among older children using different indicators including more detailed quantified measures of intake. A study among older preschool aged children (36-59 months of age) in Burkina Faso using 24 hour recall methods found slightly greater dietary diversity in the post-harvest season (3.88/7 food groups) vs. the lean season (3.64 food groups), despite finding fairly notable differences in nutrient adequacy of 10 nutrients (0.52 vs. 0.43).(31) Given the study population consisted of
older children, breastmilk was not included, and it is possible that inclusion of breastmilk could attenuate seasonal effects in an 8 item scale, particularly in rural settings with high breastfeeding rates. In a population of school aged children 4-8 years of age in rural Zambia, researchers found that seasonality was predictive of a 7-item score and also influenced the ability of the dietary diversity score to capture nutrient adequacy. In rural Tanzania, researchers found no difference in a 1-day or 7-day mean dietary diversity scores among a population of 2-5 year old children, but found a slight difference in 1-day food variety scores, driven by the use of wild/forest foods during the wet food insecure season.

Lack of seasonality of the FGS and certain components (i.e. specific food groups) could be explained in part by a “buffering effect” as different items in a food group become available across different seasons. For example, provitamin-A rich orange fruits such as mangoes often have widespread availability during a very limited season in many countries. In some cultures, fruits are also perceived to be more of a child’s food which could strengthen seasonal patterns of consumption in this age group. During seasons of food shortage, households may also buffer the effects of seasonality by preferentially allocating food to children, as observed in a study from Benin. It is also possible that items within a food group may change over time, minimizing variability around the year; for example green leafy vegetables are grouped into the same category as orange-fleshed fruits, and so lack of availability of orange fleshed fruits in some seasons could be buffered by greater availability of green leafy vegetables. Wild greens are often locally available across the year in the food system and poorer households may rely on them more. Such effects could help to explain why for most food groups we saw little fluctuation by month in the contexts that we examined. Because flesh foods are seldom consumed in many contexts, it would be expected that variation is largely random. However, flesh food consumption could be higher in festival seasons which vary by country in terms of the consistency of their timing. Breastmilk may be a stable source of food across seasons, although there is some evidence of slight seasonality to consumption patterns. There was some evidence that inversely correlated seasonal variation across different food groups may act to cancel each other out, in Senegal in particular, and moreover that breastmilk may act as a buffer for staples in particular in both Nepal and Senegal.
Other plausible explanations exist for the lack of strong seasonal patterns observed for MDD and FGS in our study. Nepal and Peru have some of the greatest variation in altitude and agro-ecological zones in the world. If different parts of the country are following different seasonal patterns, there could be a smoothing of seasonal effects at the national level. Our observation that seasonal patterns were much stronger in rural than urban areas lends support to this argument. Another plausible explanation for lack of strong seasonality is that the indicator simply measures consumption of each item within the past day, not the amount consumed. If foods are consumed in lesser amounts during the lean season, but still consumed, as observed for meat and fish in Benin (34), dietary diversity indicators may not be responsive to such changes. If detecting seasonal changes over time is a measurement priority, the use of measurement approaches that incorporate quantities of foods consumed could be important.

Our study also revealed that diets of children living in urban areas of both Peru and Senegal are more diverse than diets of children in rural areas of both countries, a finding that is consistent with the literature (36). However, because urban populations tend to purchase rather than grow their foods, access to healthier (and more seasonal) foods such as fruits and vegetables tends to be strongly tied to socioeconomic status. While we did not detect the presence of seasonal effects on the diets of children in urban Peru or Senegal, further work might explore whether such effects might exist within specific socioeconomic subgroups of urban children.

An important limitation to our study was that the Nepal sample was much smaller than the Peru and Senegal samples. It was also drawn from a smaller geographical area (nine wards total from three zones vs. national samples, limiting the external generalizability of findings. The small sample size resulted in wide confidence intervals around estimates and likely underpowered tests of statistical significance for seasonal effects. However multiple observations taken from the same children, households, and areas over time in Nepal and the near-complete enumeration of these areas allowed for us to better control for external sources of variation and to minimize sampling errors.

In conclusion, we observed the presence of seasonal patterns in MDD in Peru and Senegal but our sample size for Nepal likely limited the ability to detect seasonality. The magnitude of seasonal effects appeared smaller than we initially expected, but was large in relation to longer term trends. If large-scale surveys were
conducted in different seasons, it could be large enough to lead to incorrect inferences about change over time. We recommend that whenever possible, DHS, MICS, and locally planned national nutrition surveys be conducted in the same season over time. When they are not, the timing of surveys relative to season should be considered when interpreting trends. While our work sheds some light on the seasonal patterns of MDD in three very different contexts, additional work is needed in other settings to explore how seasonality influences this and other dietary diversity indicators (2). Further investigation is also needed to understand the potential role of region-season interactions in attenuating seasonal shifts in national estimates of MDD of Peru and Senegal.

Acknowledgements

We are very grateful for the participation of PoSHAN study respondents. We thank Professor Keith P. West Jr. for leadership of the PoSHAN study and Riley Auer support with the literature review for this study. We also acknowledge the significant contributions of the New ERA and Nepali Technical Assistance Group data collection teams under the leadership of Jagat Basnet, Kshitiz Shrestha, Sidhartha Tuladhar, Deepak Thapa, the Nutrition Innovation Lab- Nepal/JHU research team: Dev Raj Gautam, Ruchita Rajbhandary, Hem Raj Paudel, Shiva Bhandari, Hari Krishna Shah, Raman Shrestha, Sudeep Shrestha, Rajan Chalise. We also acknowledge the support from the Government of Nepal for the parent study and the leadership and partnership of Tufts University.

Author contributions are as follows: ALTL, RH, LEMB designed the research. BS, AKC, SM provided essential materials (curated datasets). SM, RDK, and BS conducted the research in Nepal. LEMB, ALTL, HK performed statistical analysis. ALTL, LEMB, RH wrote the paper. ALTL has primary responsibility for final content. All authors have read, given feedback on, and approved the final manuscript.
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Table 1. Characteristics of the datasets and populations

|                                | Peru    | Senegal | Nepal¹ |
|--------------------------------|---------|---------|--------|
| # children 6-23 months of age  | 35,272  | 12,183  | 1,309  |
| Rural, %                       | 32.0    | 64.6    | 100    |
| Food group score,¹ Mean, (sd)  | 5.4 (1.87)| 2.9 (1.54)| 4.09 (0.29)|
| Minimum dietary diversity, % ≥5| 75.4    | 15.7    | 39.1   |
| Height for age z score, mean, (sd) | -1.04 (1.014)| -0.89 (1.19)| -1.75 (0.08)|
| Stunting (HAZ < -2 z scores) (%) | 16.4    | 15.9    | 41.1   |
| Weight-for-height z score, mean, (sd) | -0.44 (1.179)| -1.12 (1.12)| -0.96 (0.13)|
| Wasting (WHZ < -2 z scores), %  | 8.0     | 21.7    | 15.4   |

All statistics are across sample children, not households. For Peru and Senegal, estimates are weighted. ¹In Nepal, the number of children represents total number of measurements conducted among the 615 unique children.
Table 2. Minimum dietary diversity and food group score by season

|        | Peru 1,2 | Senegal 1,2 | Nepal 1,3 |
|--------|----------|-------------|-----------|
|        | MDD      | FGS         | MDD       | FGS       | MDD      | FGS       |
| N=35,272 | N=35,272 | n=12,183    | N=12,183  | n=1,309   | n=1,309  |
| Lean season | MDD  | FGS         | MDD  | FGS         | MDD  | FGS         |
| -0.0179** | -0.087** | [-0.033, -0.002] | [-0.154, -0.021] | -0.0347*** | -0.140*** | [-0.058, -0.011] | [-0.245, -0.035] |
| Post-harvest | -     | -           | -     | -           | -0.0139 | -0.123    |
| -       | -0.004   | 0.004       | 0.006  | 0.005       | [-0.130, 0.102] | [-0.467, 0.222] |
| R²      | 0.004    | 0.004       | 0.006  | 0.005       | 0.010   | 0.019      |
| Joint F-test p-value | 0.006 | 0.005 | 0.010 | 0.019 | 0.938 | 0.692 |

1 All models fit using ordinary least squares (OLS); year fixed effects included in all models.
2 Standard errors are clustered by DHS cluster, table presents 95% confidence intervals.
3 Nepal fit with simple OLS with standard errors clustered by cluster, lean season represents monsoon lean season. Monsoon lean and postharvest season are tested against omitted winter lean season and the F-test p-value at the bottom of the table tests the joint significance of both seasons.

*p<0.1, **p<0.05, ***p<0.01
Table 3. Minimum dietary diversity and food group score by season and rural/urban locality in Peru and Senegal

|                  | Peru |                  | Senegal |
|------------------|------|------------------|---------|
|                  | MDD  | FGS              | MDD     | FGS     |
| Lean season*rural|      |                  |         |         |
| N=35,272         |      |                  | n=12,183|         |
| -0.0296**        | -0.126**       | -0.0309**       | -0.127**|
| [-0.055, -0.004] | [-0.228, -0.024] | [-0.054, -0.008] | [-0.230, -0.024] |
| Lean season*urban|      |                  |         |         |
| -0.000210        | -0.0193        | -0.0298         | -0.103  |
| [-0.018, 0.018]  | [-0.099, 0.061] | [-0.082, 0.022] | [-0.324, 0.118] |
| Urban            |      |                  |         |         |
| 0.134***         | 0.535***       | 0.0971***       | 0.486***|
| [0.117, 0.151]   | [0.465, 0.604]  | [0.066, 0.128]  | [0.358, 0.614] |
| R²               | 0.028          | 0.023           | 0.022   | 0.028   |

1 All models fit using ordinary least squares (OLS); year fixed effects included in all models.
2 Standard errors are clustered by DHS cluster. Table presents 95% confidence intervals.
*p<0.1, **p<0.05, ***p<0.01
Figure 1. Prevalence of MDD and mean FGS by survey year in Peru and Senegal
Figure 2. Prevalence of MDD and mean FGS by month for Peru, Senegal, and Nepal, data pooled across survey years
Figure 3. Prevalence of MDD and mean FGS disaggregated by lean season in Peru, Senegal and Nepal
Figure 4. Prevalence of MDD and FGS disaggregated by lean season and rural/urban locality, Peru and Senegal
Figure 5. Prevalence of consumption of specific MDD components by month for Peru, Senegal and Nepal