

1 Simple methods to obtain a food listing and portion size distribution estimates for
2 use in semi-quantitative dietary assessment methods

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4 Food listing and portion size estimation methods for dietary assessment

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15

17 **Abstract**

18 Semi-quantitative dietary assessment methods are frequently used in low income countries, and
19 the use of photographic series for portion size estimation is gaining popularity. However, when
20 adequate data on commonly consumed foods and portion sizes are not available to design these
21 tools, alternative data sources are needed. This study aimed to develop and test methods to: (i)
22 identify foods likely to be consumed in a study population in rural Uganda, and; (ii) to derive
23 distributions of portion sizes for common foods and dishes. A process was designed to derive
24 detailed food and recipe lists using guided group interviews with women from the survey
25 population, including a ranking for the likelihood of foods being consumed. A rapid recall
26 method to estimate portion sizes using direct weight by a representative sample of the survey
27 population was designed and implemented. Results were compared to data from a 24 hour
28 dietary recall. Of the 82 food items reported in the 24 hour recall survey, 87% were among those
29 ranked with a high or medium likelihood of being consumed and accounted for 95% of
30 kilocalories. Of the most frequently reported foods in the 24 hour recall, portion sizes for many
31 (15/25), but not all foods did not differ significantly ($p < 0.05$) from those in the portion size
32 estimation method. The percent of portion sizes reported in the 24 hour recall between the 5th
33 and 95th percentiles determined by the portion size distribution estimation method ranged from a
34 low of 18% up to 100%. In conclusion, a simple food listing and ranking method effectively
35 identified foods most likely to occur in a dietary survey. A simple method to obtain reliable
36 portion size distributions was effective for many foods, while the approach for others should be
37 modified. These methods are an improvement on those in current use.

39 **Introduction**

40 Dietary assessment surveys are necessary to adequately inform, design and evaluate nutrition intervention
41 programs in low-income countries. While the 24-hour dietary recall (24HR) is one of the most common
42 method used in these settings [1], it is also more resource intensive and technically challenging than other
43 methods [2,3] and this may be a limiting factor for the use of dietary surveys to inform effective nutrition
44 programs. Depending on the specific objectives of a dietary survey, simpler and less resource intensive
45 methods, including food frequency questionnaires (FFQ) or semi-quantitative (SQ) FFQ, may be
46 adequate [4]. The semi-quantitative estimation of portion sizes consumed using food photo series or
47 atlases depicting graduated portion sizes for a variety of foods is gaining popularity, and may be used to
48 support the application of SQ-FFQs [5] or as a way of simplifying 24HR methods [6,7].

49 However, to adequately develop such SQ dietary assessment tools, some key information is required
50 beforehand. This includes, but is not limited to: (i) a listing of the foods that are commonly consumed in
51 the study population and hence should be included in the SQ tools; (ii) relevant details about the way they
52 are typically prepared or consumed, and; (iii) the distribution of usual portion sizes to select those to
53 represent in the SQ tools. Ideally, data-driven methods in the form of previously collected dietary intake
54 survey data that is quantitative, valid, and representing the same survey population and sub-population
55 groups of interest, would serve this purpose [5,8]. In low-income countries where data meeting these
56 criteria may often not be available, some form of reliable, empirically-derived preliminary data are
57 needed.

58 We have found few well-described or well-designed processes in the published literature on how to
59 collect food listing and portion size distribution data when appropriate previous survey data are not
60 available. Food listings have been derived using informal or subjective methods such as consultation with
61 food service professionals, local cook books, or restaurant and cafeteria menus, or conducting interviews
62 with cooks or chefs in households and restaurants, but without any information on the sample size,

63 sampling frame or representativeness [6,9,10]. While these are relatively low-cost methods, it is not clear
64 how complete or representative they are.

65 In the absence of pre-existing data, portion size ranges have been deduced by various means, such as by
66 adapting from local reference data (e.g., dietary guidelines and nationally established standard serving
67 sizes), consulting experts in the catering industry, qualitative consultation with households [9-11], or
68 deriving a medium portion size from existing survey data but applying fixed ratios to derive small and
69 large portion sizes [9,12]. In some studies to develop and validate food photo atlases, either very scant or
70 no information was provided on how graduated portion sizes were derived [10,13,14], confirming that
71 this methodological step is often overlooked. This is concerning as, if portion size options presented in
72 SQ-FFQ or SQ-24HR dietary assessment surveys do not represent the usual range consumed in the
73 population studied, large portion size estimation errors can result [15].

74 In some studies, mean portion sizes were determined in small surveys where householders were asked to
75 demonstrate usual portion sizes for different foods for specific age groups [6,12,16], but a minimal
76 description of the methods used for sampling or data collection was provided. This may be an innovative
77 way to collect portion size data for dietary survey tool design, but well-described methods using a
78 systematic and representative approach are needed.

79 The aim of this study was to develop, document and field test data collection methods to determine food
80 listings and portion size distributions for application in dietary assessment studies. We chose to conduct
81 this work among women in rural Uganda, where researchers have experience in conducting large-scale
82 24-HR dietary recall and SQ-FFQ surveys. The main objectives were to develop, document, and field test
83 methods to: (i) create listings of foods likely to be consumed in a study population, and; (ii) to
84 quantitatively derive distributions of portion sizes for commonly consumed foods and composite dishes.

85

86 **Materials and methods**

87 This study was conducted as part of a larger study to compare dietary intake outcomes of a SQ
88 24-HR survey method and a SQ-FFQ method with a standard 24-HR method (to be reported
89 elsewhere). The food listing method used a qualitative approach with categorical ranking of the
90 likelihood of foods being consumed, while the method to derive portion size distributions used a
91 quantitative recall approach. The sampling frame was established for the larger study and was
92 used to draw sub-groups of participants for the data collection methods described here. The study
93 was conducted in Nakisunga sub-county (population >48,000) in Mukono District, Uganda, a
94 site that was purposively selected for its proximity to Kampala, having a largely rural
95 agricultural livelihood with some urban influence, socio-cultural homogeneity, and the
96 cooperation of local authorities.

97 This study was reviewed and approved by the Higher Degrees, Research Ethics Committee,
98 Makerere University School of Health Sciences, Kampala, Uganda and registered and approved
99 by the Uganda National Council of Science and Technology. Informed written consent was
100 obtained from all participants.

101

102 **Study participants and sampling**

103 The study included women 18-49 years of age, who were primary residents of the home visited,
104 self-identified as the primary or most senior female caretaker in the household with
105 responsibility for meal preparation, and were available and consented to participate. Women
106 who self-reported to be currently pregnant or lactating with a child <23 months of age were
107 excluded.

108 We used a multi-stage sampling procedure whereby four of eight parishes in Nakisunga sub-
109 county were randomly selected. Three enumeration areas (EAs), defined by the 2014 Uganda
110 Population and Housing Census sampling frame, were randomly selected from each of the four
111 parishes. A household census of these EAs identified households with eligible women. After
112 dividing the total sample equally among the four parishes, population proportionate sampling of
113 eligible women was done from the three selected EAs for each of the data collection activities.
114 Based on the sample size calculation for the larger study, the samples for these studies were
115 drawn from a pool of n=336 women.

116

117 **Data collection**

118 **Socioeconomic data**

119 A brief socioeconomic questionnaire was administered to all participants of the larger study. We
120 used the Progress out of Poverty Index® (PPI), as validated in Uganda, to compare poverty risk
121 among participants in different data collection activities. The questions, indicators and scoring
122 methods were downloaded from the PPI website
123 (<http://www.progressoutofpoverty.org/country/uganda>).

124 **Food listing**

125 The food listing activity was designed to create a list of all foods and beverages commonly
126 consumed by the survey population, including nutritionally-relevant details such as state (e.g.,
127 raw/cooked; ripe/unripe), processing method (e.g., dried, milled/extracted, fermented), or
128 cooking method, and to capture main and optional ingredients of commonly consumed recipes.

129 Two types of semi-structured interviews were used to elicit these details; one with key
130 informants (KI) and the other with groups of survey participants. Prior to the interviews, project
131 staff created a spread sheet with food group categories and an initial list of foods likely to appear.
132 Two KI interviews were held, one with district level and one with sub-county level government
133 staff. Each interview included a government agriculture officer and a health officer, selected for
134 their knowledge of the availability of local foods, diets, and seasonal availability. A field
135 coordinator and field staff member conducted the interview using the initial food group/food
136 item list as a guide and as a prompt list for foods not mentioned by the KIs, while another field
137 staff member recorded the information. The purpose of the interview, the information of interest,
138 and the ranking categories for likelihood of food items being consumed at the time of the
139 prospective dietary survey (i.e., July 2017) were explained. Detailed recipe data were not
140 obtained in the KI interviews. Data collected from the two interviews were transcribed into a
141 spreadsheet format and combined, and average rankings were calculated, always rounding to the
142 higher likelihood ranking when rankings did not conform. An expanded listing of available food
143 items and common dishes, an average ranking of the of food item availability, and the main
144 processing methods, was derived. This listing was then used to guide the group interviews.
145 Four guided group interviews (GGI) were held with women randomly selected from the sample
146 list (8-10 women from each of 4 parishes), where each interview covered half of the food groups
147 and lasted approximately 2.5-3 hours. A structured guide was used to elicit food items
148 consumed, including specific details on the food type (e.g., local name(s), color, variety,
149 commercial products), processing and preparation methods (e.g., whole or milled; mashed or
150 chopped and boiled, steamed, fried, etc.), the likelihood of the food being consumed in the
151 household during the survey period (i.e., high, medium, low, not likely at all), and recipes for

152 mixed dishes prepared. For each mixed dish recipe type mentioned, additional details were
153 obtained, including preparation method, whether ingredients were major or minor components,
154 and a likelihood ranking for their inclusion in the recipe (i.e., 'always', 'often', or 'rare'; this
155 information is used to correctly identify recipes and for the purpose of collecting standard recipe
156 data, a process not reported on here). Finally, a listing of the most common ingredient
157 combinations was obtained. The data collection tools with completed examples used to record
158 and summarize responses for food items/ingredients and mixed dish recipes, with sample data,
159 are given in S1 and S2 Figs, respectively.

160 Information from the two interviews on the same food groups was combined in a spreadsheet
161 (one for food items and one for recipe data), and an average ranking score of foods or ingredients
162 was obtained, rounding to the higher frequency category.

163 **Selection of foods for portion size distribution estimation (PSDE)**

164 Foods consumed as individual items (i.e., not as ingredients in mixed dishes) that were ranked
165 with a high or medium likelihood of being consumed (n=43), plus mixed dishes made with those
166 foods (n=24), were included in the PSDE method, for a total of 67 foods. For the selection of
167 individual food items, this included the 43 high or medium ranked foods identified in the food
168 listing interviews, plus 3 that had oil-fried versions and were distinguished, plus 7 processed
169 baked goods items that were not well addressed in the interviews but added by the researchers as
170 they were considered common in the area. From this total of 53 individual food items, 10 were
171 dropped, as one was not found in the market (i.e., apples), one was better estimated as a count
172 than portion size distribution in grams (i.e., hard candies), and 8 were similar to other items and
173 the portion size was not expected to differ between them (i.e., different meat types, and different
174 varieties of sweet potato, amaranth leaves, yams and some bananas). For the mixed dishes, 30

175 common ones had been identified, but 6 were dropped as portion sizes were expected to be the
176 same for very similar mixed dishes. For some mixed dishes for which primary ingredients are
177 substitutable, a mixed dish ‘type’ was used to represent the variations (e.g., dishes made with
178 similar types of green leaves or common beans were grouped together).

179 **PSDE method**

180 Usual portion sizes for different foods were determined with participants using interactive
181 interviews. The 67 selected foods were divided into 4 sets of 16-17 items each, and portion size
182 estimation data were collected on separate days for each set. Four subgroups of 56 women
183 (n=224) invited to participate were asked to recall portion sizes for one of the four sets of foods.
184 We calculated sample sizes for a range of different foods using existing portion size data from a
185 dietary survey conducted in central and eastern Uganda using the equation: $[Z\alpha/2 \cdot \delta / E]^2$,
186 where $Z\alpha/2 = 1.96 = 95\%$ confidence, $\delta =$ known SD and $E =$ acceptable error in measurement
187 units. The error (E) was set at the equivalent of a coefficient of variation of 15%. This resulted in
188 sample sizes ranging from n=13 to 135, and 80% of the 15 sample sizes calculated were n<60.
189 We rationalized that n=56 data points would be adequate for most foods.

190 The portion size estimation sessions were organized in a central location of each parish. All
191 foods and dishes were prepared by locally hired assistants in the form typically served. The sets
192 of 16-17 foods were divided into 3 separate data collection stations, with one interviewer and one
193 person weighing and recording the portion sizes, with each woman completing data collection at
194 one station before moving to the next one. For each food item, the interviewer prompted the
195 woman to recall if that food was consumed on the previous day, week, or months. If they could
196 not recall the last time they ate that food, or they never eat that food, no information was
197 collected. If they could recall the last time they ate that food, they were asked to estimate the

198 amount consumed. The respondent was then asked to serve up that amount of food from the real
199 foods provided. These amounts were weighed to the nearest gram on a digital dietary scale,
200 recorded, and the weight of the dish subtracted.

201 Where portion size data were collected with inedible fractions included (e.g., bones in fish, peel
202 and seeds in watermelon), this was also recorded. Edible fractions for those foods was
203 determined separately by weighing a sample of food items, removing inedible fractions, and then
204 weighing the yield of edible amount on dietary scales. Edible yield factors were then applied to
205 the portion sizes recorded to calculate the weight of the edible portion size.

206 **24-HR Survey**

207 We used a multiple pass approach based on Gibson and Ferguson [17] with specific methods that
208 were previously described in detail [18]. Group 'training' sessions were held in each EA two days
209 before the 24HR interview to explain the purpose of the study, and the methods involved. They
210 were asked to use their own dishes for serving and eating their food the next day to improve
211 visual memory, and instructed on the use of picture charts to mark foods consumed.

212 Portion sizes of items consumed were estimated using methods specified for each food type.
213 These included life-sized graduated photographs, weighing scales, graduated measuring
214 cylinders and play dough models, or standard weights for foods that are served as units (e.g.,
215 boiled egg, bread slice) [17]. Portion sizes recorded accounted for any leftovers that were served
216 but not consumed. If multiple servings of the same food item were reported to be consumed in a
217 single eating occasion (e.g., morning, afternoon, or evening meals or snacks) these amounts were
218 combined to a single portion. All of these proxy measures were later converted to gram weights
219 of the food represented using a set of conversion factors.

220 **Data management and analysis**

221 The CSDietary program (HarvestPlus/Serpro, 2009), using the CPro software platform (Serpro,
222 Santiago, Chile), was used for dietary data entry and data processing. All data were entered in
223 duplicate and discrepancies were identified and rectified and distributions of intakes were
224 reviewed for plausibility by examining high and low intakes. All subsequent data management,
225 processing, and analyses were done using Excel (Microsoft Office 2007 for Windows) and SPSS
226 16.0 and 18.0 for Windows (SPSS Inc., Chicago, IL, USA). For the food listing, the number of
227 foods by likelihood ranking was determined. For the portion size estimation activity, and portion
228 sizes derived from the 24HR survey, descriptive statistics (mean, SD, CV, and 5th, 50th, and
229 95th percentiles) were calculated for the portion size distributions in grams.

230 To determine whether the distribution of portion sizes derived from the PSDE method were
231 adequate to capture those reported in the 24HR survey, we: (i) calculated the number (percent) of
232 portions reported in the 24HR survey whose gram weights fell within the 5th and 95th
233 percentiles of weight derived in the PSDE activity. These percentiles are suggested to represent
234 the smallest and largest portion sizes in food photo series [8], and; (ii) compared the distributions
235 using the Mann-Whitney U-Test/Wilcoxon Rank Sum Test for two independent samples with
236 unequal sample sizes drawn from the same population, where $p < 0.05$ indicates a statistically
237 significant difference. For the socio-demographic data, each individual indicator or score was
238 compared between the PSDE method group and the 24HR survey group, as was the final PPI
239 score.

240

241 **Results**

242 **Sample and socio-demographic data**

243 The participation rate in the PSDE was 96% (i.e., 214/224). Socio-demographic data were
 244 derived for only a subset of 86% (184/214) of the PSDE participants as these data were collected
 245 only for those who also participated in a larger study, including the 24HR survey presented here
 246 (Table 1). Of the 184 PSDE participants, 57 participated in both the PSDE and the 24HR survey
 247 and the data were retained in both groups for this analysis. Results for these subgroups suggest
 248 that they were similar in socio-demographic characteristics (Table 1), except (Table 1).

249

250 **Table 1. Socio-demographic data for subgroups of participants in the PSDE and 24HR**
 251 **surveys^a**

		PSDE method a		24HR survey ^a		
Characteristic	Response	Mean	SD	Mean	SD	P ^b
n		184 ^c		115		
Age (years)		33.6	9.0	33.4	9.0	ns
Number of household members (n)		5.9	2.5	5.7	2.5	ns
		%	CI	%	CI	
All household members own at least one pair of	Yes	81.0	75.3 - 86.7	76.5	64.7 - 81.3	ns

shoes (%)						
All children 6-12 years in school (%)	Yes	67.9	61.2 - 74.6	64.3	55.5 - 73.1	ns
	No children 6-12 years	4.9	1.8 - 8.0	0.9	0.0 - 2.6	ns
Lead female able to read/write (%)	Yes	77.2	71.1 - 83.3	79.1	71.7 - 86.5	ns
Main wall material (%) ^d	Brick, earth or clay	92.4	88.6 - 96.2	91.3	86.1 - 96.5	ns
Main roof material (%) ^d	Iron sheets	97.8	95.7-99.9	96.5	93.1 - 99.9	ns
Toilet facility type (%) ^d	Pit latrine with cement slab	58.7	51.6-65.8	54.8	45.7 - 63.0	ns
	Pit latrine - no cement slab	20.1	14.3 - 25.9	23.5	15.8 - 31.2	
Cooking fuel type (%) ^d	Wood / dung / grass	58.2	51.0 - 65.3	55.7	46.6 - 64.8	ns
	Coal	41.8	44.7 - 49.0	44.3	35.2 - 53.4	
Number of cell phones (%)	0	7.6	3.8 - 11.4	2.6	0 - 5.5	ns
	1	22.3	16.3 - 28.3	21.7	14.2 - 29.2	
	2	44.0	36.8 - 51.2	47.8	38.7 - 56.9	
	≥3	26.1	19.7 - 32.3	27.8	19.6 - 36.0	
PPI score ^a		53.6		54.7		

252 ^aPSDE, Portion size distribution estimation method; 24HR, 24 hour dietary recall; PPI, Progress
253 out of Poverty.

254 ^bANOVA test where means are presented and by Chi-square test where data are categorical; *,
255 $P < 0.05$; ns, non-significant ($P \geq 0.05$).

256 ^cData on socio-demographic data are available for only 184 of the 214 participants in the PSDE
257 data collection as this questionnaire was only applied to those who participated in a household
258 survey including three dietary assessment methods. Of the 184 PSDE participants, 57 also
259 participated in the 24HR survey presented here and the data were retained in both groups for this
260 analysis.

261 ^dData only shown for the primary responses recorded; statistical tests included all possible
262 responses.

263

264

265 **Food listing**

266 The food and recipe listing process identified 77 unique foods (i.e., those consumed as individual
267 food items and those used as ingredients in composite dishes) that were ranked in the GGI with a
268 high or medium likelihood of being consumed during the survey, including 3 foods with two
269 distinct preparation methods. Likewise, 48 were ranked with a low likelihood, and 57 as not at
270 all likely or not consumed at all.

271 Of the 82 distinct foods and ingredients mentioned in the 24HR survey, 71 (87%) were among
272 those ranked with a high or medium likelihood of being consumed and accounted for 95% of

273 estimated kilocalorie intake, while 7% were among those ranked with a low likelihood or not at
274 all likely, accounting for <1% of estimated kilocalories; of the latter, 5 foods were reported by a
275 single individual and 1 food was reported by 2 individuals. The remaining 6% of foods reported
276 in the 24HR did not appear in the food listing at any stage, and accounted for 5% of estimated
277 kilocalorie intake; of these, sugarcane was reportedly consumed by 26 individuals, while the 4
278 other foods were reported by ≤ 4 individuals.

279 The KI food listing tended to result in higher rankings of foods than the FGDs. For example,
280 there were 24 foods ranked as not likely or never consumed by the FGDs that were ranked with a
281 high (n=1), medium (n=8) or low (n=15) likelihood of being consumed. There were also 10
282 foods listed as being consumed by the KIs but not mentioned or ranked during the GGI. The
283 latter were largely comprised of uncommon bean varieties and non-indigenous vegetables.

284

285 **PSDE Method**

286 Descriptive data are presented for the distributions of estimated portion sizes for a selection of
287 individual food items and mixed dishes representing those reported with highest frequency (i.e.,
288 >10 occurrences; Table 2) and with low frequency (i.e., 4-6 occurrences; Table 3) in the 24HR
289 survey. The SDs for these food items were relatively large, and the coefficient of variation (CV)
290 for these estimates ranged from 0.27 to 0.98, with an average of 0.47. Portion sizes for
291 approximately half of the individual food items and mixed dishes did not follow a normal
292 distribution ($p < 0.05$).

293

294 **24HR survey**

295 Only 18 individual food items and 11 mixed dishes had ≥ 10 reported portions consumed,
296 including multiple portions consumed by the same person on the day of recall. Descriptive data
297 (Tables 2 and 3) are not given for 4 individual food items as these were not considered in the
298 portion size estimation activity; two had similar substitute foods included, one was to use a
299 standard unit size as the basis for portion size estimation so was excluded and one food was not
300 picked up in the food listing exercise. For those reported, the SDs were also relatively large and
301 the CVs ranged from 0.24 to 0.93, with an average of 0.49.

302 **Table 2. Portion sizes for foods and composite dishes estimated from a portion size recall survey and reported with frequency**
 303 **≥10% of all food portions in a 24HR survey in the same population**

	PSDE Method							24HR Survey					
Food or Beverage	n	Mean	SD	CV	5th pct ¹	Median	95th pct	n	Mean	SD	CV	Median	Portions within 5 th -95 th pct (%) ^b
Individual food items													
Plantain, cooked	54	361	140	0.39	111	369	624	48	485	238	0.49	474	69*
Maize on cob, cooked	53	165	83	0.50	36	139	366	43	163	85	0.52	129	98
Cassava, boiled	50	179	87	0.49	63	179	341	42	225	133	0.59	197	76
Avocado	40	77	39	0.51	23	68	169	42	77	36	0.47	89	95
Mango	43	203	91	0.45	94	180	386	34	98	73	0.74	82	18*
Sweet potato, yellow or white, cooked	49	327	162	0.50	102	327	655	56	322	198	0.61	283	86
Bread, white	58	78	33	0.42	25	78	141	28	160	94	0.59	125	57*

Chapatti	50	129	50	0.39	77	111	252	22	93	37	0.40	94	68*
Banana, large-type	53	113	70	0.62	45	120	215	20	117	66	0.56	104	90
Mandazi (fritters)	47	98	95	0.97	41	91	132	18	63	37	0.59	63	67*
Beef, cooked	53	50	28	0.56	20	43	101	15	102	64	0.63	99	53*
Jackfruit	56	305	174	0.57	124	295	733	13	316	190	0.60	276	100
Pumpkin, cooked	53	231	142	0.61	54	212	541	13	197	63	0.32	190	100
Egg, fried	47	57	28	0.49	27	47	107	11	91	33	0.36	99	82
Mixed dishes													
Bean sauce	51	236	73	0.31	122	220	372	76	191	102	0.53	187	66*
Maize posho (stiff porridge)	49	328	142	0.43	98	340	549	52	329	127	0.39	326	94
Milk tea	55	430	115	0.27	250	398	579	47	422	152	0.36	390	89
Rice dish	49	285	104	0.36	88	301	463	43	297	171	0.58	245	79

Small fish sauce	53	50	20	0.40	19	47	87	24	133	59	0.44	136	21*
Fresh fish soup (broth only)	46	72	38	0.53	29	64	163	19	162	86	0.53	146	58*
Cassava and beans (Katogo)	49	458	163	0.36	177	436	730	17	419	232	0.55	367	82
Fruit juice, fresh (single or mixed)	51	315	96	0.30	160	296	482	14	334	95	0.28	312	71
Eggplant/entula sauce	48	194	60	0.31	95	182	325	13	152	69	0.45	136	77
Beef soup (broth only)	53	73	38	0.52	12	70	141	11	148	78	0.53	136	55*
Maize porridge, refined flour	55	540	227	0.42	293	547	1134	11	470	115	0.24	461	100

304 ^aPSDE, Portion size distribution estimation method; 24HR, 24 hour dietary recall, PCT, percentile.

305 ^bThe percentage of portions for individual foods or mixed dishes with portion sizes (grams) falling within the 5th-95th percentile range
306 of portion sizes derived by the PSDE method. * indicates statistically significant differences between portion sizes (grams) between
307 the PSDE method and the 24HR survey data; Mann-Whitney U-Test, P < 0.05.

308 **Table 3. Portion sizes for selected foods and composite dishes estimated from a PSDE survey and reported with low frequency**
 309 **(i.e., 4-6% of all food portions) in a 24HR survey in the same population^a**

Data Source	PSDE Method							24HR Survey					
Food or Beverage	n	Mean	SD	CV	5th pct	Median	95th pct	n	Mean	SD	CV	Median	Portions within 5 th - 95 th pct (%) ^b
Individual food items													
Chips (French fried potatoes)	40	144	65	0.45	56	140	261	5	227	94	0.41	260	60
Groundnuts, roasted	50	51	24	0.47	20	48	100	5	67	35	0.52	67	67
Fish (large species), dried, boiled	49	104	40	0.38	43	93	196	5	27	25	0.93	21	20
Banana, small type	41	148	40	0.28	92	142	220	5	160	124	0.78	114	20
Samosa	49	64	63	0.98	12	61	191	5	95	45	0.47	94	100

Mixed dishes													
Plantain & beans (katogo)	48	463	154	0.33	260	440	764	6	434	136	0.31	432	100

310 ^aPSDE, Portion size distribution estimation method; 24HR, 24 hour dietary recall, PCT, percentile.

311 ^bThe percentage of portions for individual foods or mixed dishes with portion sizes (grams) falling within the 5th-95th percentile range

312 of portion sizes derived by the PSDE method.

Comparison of portion sizes between the PSE activity and 24HR survey

Of the foods reported with relatively high frequency in the 24HR survey (Table 2), the median portion sizes for many (15/25), but not all foods, were not significantly different from those determined in the PSDE method. For foods with medians that differed significantly, there was no systematic bias in the direction of difference. The percent of portion sizes reported in the 24HR survey that fell between the 5th and 95th percentiles determined by the PSDE method ranged from a low of 18% up to 100%. Of the foods reported with lower frequency in the 24HR survey (Table 3), results were similar. The percentage of portion sizes falling between the 5th and 95th percentiles differed markedly among the 6 food items shown here, ranging from 20 to 100%.

Discussion

We have described two relatively low cost methods that could aid the development of semi-quantitative dietary assessment methods, as determined in a rural African population. A simple food and recipe listing method found that a ranking system was very effective at identifying foods that were most likely to occur in a dietary survey. A relatively simple and rapid method to obtain reliable distributions of portion sizes from a minimum sample indicated that for many foods, portion size distributions compared well with those obtained from standard 24HR methods, while several others did not.

The food listing method developed and field-tested here provides a useful, categorical method to identify foods that should be included in a food list for dietary surveys using closed lists, such as

FFQ and SQ-FFQ methods. The foods identified by the ranking process as having a high or medium likelihood of being consumed covered the vast majority (i.e., 95%) of the total kilocalorie intake in the 24HR survey. This is important as it is a key criteria for developing adequate FFQ/SQ-FFQ methods [5]. If the foods that were ranked here as having a low likelihood or unlikelihood of being consumed were omitted from a FFQ or SQ-FFQ survey derived from it, this would have accounted for a negligible proportion of kilocalories being missed by the survey (i.e., <1%). Although this process may be best suited to general surveys that aim to assess intakes from all foods, it could easily be adapted for use with specific food groups or foods providing specific nutrients.

In addition to use in FFQ and SQ-FFQ surveys, this food listing process could be used to support SQ-24HR methods such as those applying food photo atlases for portion size estimation. It would also be recommended to prepare for standard 24HR surveys as it allows survey designers to create prompt lists for relevant food details that should be probed for in an interview, and to predetermine the most appropriate portion size estimation method for each food likely to occur. This is expected to enhance the training and preparation of enumerators, and possibly the quality of data collected. Very little detail or specific guidance has been provided in the literature where such food listing processes are mentioned [6,9,10,19] or recommended [20].

This food listing method would be improved by including separate likelihood rankings for foods consumed in different forms, including individual foods consumed in raw or cooked forms, foods cooked with or without oil, or as ingredients in mixed dishes so that these can be distinguished for inclusion in the survey. A small number of food items (n=5) occurring in the 24HR survey were not captured by the food listing method. These were primarily low frequency foods, but one food, sugarcane, was reported by a large percentage (i.e., 23%) of individuals.

Baked goods and some commercial beverages were also not adequately probed for during the KIs or GGIs so more careful listing and probing, particularly of processed foods or snacks, is needed, as these can be easily missed under standard food group headings. Finally, the GGI interviews were more relevant to the process as they focused on foods actually consumed in households, rather than focusing more on availability in the community as for the KIs. The latter resulted in more foods being mentioned by KIs and foods being ranked with greater likelihood of consumption than in the GGIs. Nonetheless, the KIs did serve to develop a more complete and locally relevant list of foods for use as a probing guide in the GGIs. The usefulness of combining expert consultation with ethnicity-specific details derived from the target population has been previously recommended [21].

We developed and field tested a novel method to derive portion size distributions for the purpose of developing low cost and simple portion size estimation tools, such as food photo atlases, for use in large-scale dietary surveys. The systematic nature of this method represents an improvement on those reported in the literature for similar use, as previous methods have been largely qualitative in nature, and/or the level of representativeness of the usual range of foods consumed by the target population is questionable [6,9,10,12,14,16].

Portion size estimation tools, including those using photographs, should reflect the range of amounts of foods typically consumed in the study population. A study among children [8] suggested that using age appropriate portion size options greatly reduced error in portion size estimation using photo series depicting portion sizes actually consumed by children (i.e., average of 7% error in weight estimation) compared to using the lower range of portion size photos derived for use with adults (i.e., 46% error) [22]. In an extensive review of FFQ methods [5], it was suggested that in the absence of existing survey data, researchers assigning on the FFQ

conduct small surveys to derive portion size data (e.g., 24HR, diet histories). However, in practice this may be impractical due to cost, time, and technical skill required and may not provide reliable distributions where samples for specific foods are less frequent. In the 24HR survey conducted in this study including 111 respondents, the majority of unique foods were reported <10 times, providing a very small sample from which to derive reliable portion size distributions. The method we tested here overcame that problem by quickly obtaining a large sample for each food deemed to be commonly consumed, as identified in the food listing activity.

This study did not aim to validate the PSDE method against a gold standard method, and the comparisons to the 24HR survey must be interpreted cautiously. The PSDE method was limited in that it relied on both short and longer term recall of portion sizes for foods consumed more than a day or week ago, and hence estimates may be distorted by memory. If some responses reflected 'usual' portion sizes, the width of distributions may be attenuated as fewer extremes might be reported. This might partially explain why at least some portion sizes reported in the 24HR for most foods were outside the 5th and 95th percentiles of the PSDE distributions.

The 24HR method used different portion size estimation tools, which included photos of small, medium and large items (e.g., vegetables or roots used as ingredients), and dry rice or play dough to estimate volumes, and each of these is then converted to edible portion amounts in grams using previously obtained conversion factors. Thus, some lack of conformity with the PSDE likely occurred due to the difference in methods and additional error that may be introduced by these conversions. In examining results for items for which the distributions were significantly different between methods, there was no apparent bias towards any one portion size estimation tool being consistently associated with low conformity. However, some foods with significantly

different distributions had two or more distinct sizes available, such as bread slices from small and large loaves, mandazi (fritters), and small and large mango varieties. In the PSDE method, these were combined into one distribution. However, it's possible that in the 24HR survey, more individuals were consuming the smaller sizes of those food items and hence the distributions were more skewed to the larger sizes. The lower conformity between methods for standard unit size items (e.g., bread slices, chapatti, mandazi) supports that basing portions on unit size with options for multiples or fractions of those units is a better approach than using continuous portion weights as derived from the PSDE [8]. In the case of beef, the PSDE method accounted for an average amount of bone as part of the beef portions measured, and it's possible that the 24HR method did not distinguish between meat with or without bone. These are issues that should be considered more deeply in establishing PSE methods and the way foods are included in the PSDE method, as relevant for a particular population.

In addition to supporting the development of FFQs, SQ-FFQ, and simplified 24HRs using food photo atlases for portion size estimation, the PSDE method presented here may also find use in nutrition research and advocacy tools that use linear programming. These methods identify foods that provide, or could provide, sufficient energy or nutrients to meet dietary requirements of a target population and require portion size estimates as input. These include Optifood, primarily used to derive food-based recommendations for optimizing diets of infants and young children [23], and the Cost of the Diet tool, an advocacy tool for estimating the cost of a nutritionally adequate diet [24]. Studies using Optifood typically use 24HR surveys to obtain input data [25,26], while the Cost of the Diet tool does not currently employ a satisfactory method for obtaining usual portion size data on which the models are based; this relatively low-cost method may provide an option to improve this tool.

Validating these methods were beyond the scope of this study. This would require a large-scale quantitative dietary intake survey in the same population with a large enough sample reporting intakes for a wide range of foods. However, we have provided a preliminary, detailed description of methods well beyond what is currently described in the published literature. This includes a limited evaluation comparing to data collected by a quantitative 24HR in the same population, as well as recommendations for improvement. We propose that these methods be further tested and validated when opportunities arise, such as in preparation for a large-scale or national dietary surveys.

Conclusions

We have identified a gap in available, well-described methods to collect data for deriving food lists and portion size distribution estimates for use in a wide range of dietary assessment methods where existing, suitable dietary intake data are not available. This preliminary evaluation of the methods described and field-tested here, employing qualitative, semi-quantitative and quantitative methods with representative sampling, is encouraging and we recommend efforts to identify the best method of estimating portion size distributions for different food types and to validate these approaches.

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Supporting information captions

S1 Fig. Example data collection sheet and guide for food listing in Guided Group Interviews (GGI)

S2 Fig. Example data collection sheet and guide for detailing recipes of mixed dishes in Guided Group Interviews (GGI). We recommend replacing the second column in the second table to record a likelihood ranking for each mixed dish mentioned, rather than obtaining ingredient information.