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Potentials and limitations of a food group-based algorithm to assess dietary nutrient intake of women in rural areas in Tanzania

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ABSTRACT

The aim of this study was to evaluate the accuracy of nutrient intake assessment with the food group-based algorithm "Calculator of Inadequate Micronutrient Intake" (CIMI) in comparison to the established nutrition software NutriSurvey. Using Food Frequency Questionnaires and 24-h dietary recalls of 1010 women from two rural districts in Tanzania, 23 relevant typical Tanzanian food groups were identified and subsequently the dietary protocols assessed via CIMI algorithm were compared by bivariate correlations and Bland-Altman analysis with the results of the NutriSurvey software (reference) and were set in relation to blood biomarkers of 666 participants. CIMI and NutriSurvey calculations regarding macro- and micronutrient intakes were similar. The Bland-Altman analyses and correlation coefficients of energy (0.931), protein (0.898), iron (0.775) and zinc (0.838) confirm the agreement of both calculations. The food group based CIMI algorithm is a practical tool to identify the inadequacy of macro- and micronutrient intake at population level.

ARTICLE HISTORY

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KEYWORDS Validation: nutrition assessment; iron; vitamin A: zinc

Introduction

their consent.

Micronutrient malnutrition is common in developing countries, often due to the consumption of monotonous staple-based diets rich in foods such as cereals (maize, millet and rice) and starchy roots (cassava, potatoes and plantains). While these foods provide adequate energy, they do not contain enough bioavailable iron, vitamin A and zinc, and other micronutrients, such as folate and vitamin B1 (thiamine) (Seal et al. 2005). Long-term consumption of a diet lacking micronutrient-dense foods results in hidden hunger: a chronic lack of vitamins and minerals, while being macronutrient- and energy-sufficient (Biesalski 2014; Gupta et al. 2017). Although hidden hunger often goes unidentified, due to lack of clinical symptoms, its consequences can be severe. Hidden hunger can lead to irreversible mental and physical problems, poor health and productivity, and is a leading cause of childhood mortality (Biesalski 2014). Early detection of hidden hunger is therefore very important, to successfully combat micronutrient deficiencies. This can be achieved by assessing micronutrient adequacy

of diets commonly consumed by at risk populations.

To identify nutrient gaps that can be used to estimate the risk for malnutrition of a certain cohort, rapid nutritional assessment tools are needed. The commonly used tools are Food Frequency Questionnaires (FFQ), Dietary Diversity scores and 24-h-recalls (Kennedy et al. 2013). With regard to FFQ and Dietary Diversity scores, the evaluation is relatively fast and uncomplicated, but the results are not suitable to calculate quantitative intakes of individual nutrients. Dietary Diversity score is only a proxy-indicator that can be used in longitudinal studies of one region (Kojima et al. 2020). 24-h-recalls, on the other hand, provide detailed information about nutrient intake and dietary patterns of the study group and give an estimate of the actual food intake of an individual as recalled from memory. The collection of the data is however often very complex and time intensive (Yuan et al. 2018). Afterwards the data are entered using a nutrition calculation software e.g. NutriSurvey, to calculate the macro- and micronutrient content of the foods. Data input in NutriSurvey

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requires the entry of each food and the amount consumed in grams, which is a time-consuming process. In addition, results of calculated intake amount of different 24-h-recalls can vary greatly from day to day or seasonally.

This study used a new method, which has been developed to collect and analyse data concerning macro- and micronutrient intake in a faster and easier way. The CIMI methodological approach to estimating inadequate intake of nutrients is based on the input of the cumulative amount of foods in specific food groups rather than individual foods. Data entry in CIMI can be done during the interview, which makes it possible for the interviewer to directly provide feedback to the participant about the individually estimated nutrient intake. This probably increases the motivation for the participant to provide correct and complete data. A study in Ethiopia could show, the CIMI-Ethiopia reduced the data collection time by 25% compared with the conventional 24-h-recalls (Bosha et al. 2019a). Furthermore, research projects in Indonesia, Ghana and Ethiopia demonstrated, that the food group-based approach of CIMI produces valid results of macro- and micronutrient intake compared to NutriSurvey (Jati et al. 2014; Bosha et al. 2019a, 2019b; Wald et al. 2019).

Therefore, this study aimed to validate the accuracy of the CIMI algorithm in comparison to NutriSurvey using socio-demographic and nutritional consumption data from a Tanzanian population (N=1010 women). The data were gathered within the two research projects Scale-N and Trans-SEC in two different agro-ecological zones.

Methods

Study design

This study included data from two cross-sectional studies within the projects Trans-SEC (Innovating Strategies to safeguard Food Security using Technology and Knowledge Transfer) and Scale-N (Scaling Up Nutrition: Implementing Potentials of Nutrition-Sensitive and Diversified Agriculture to Increase Food Security). The Trans-SEC study was conducted from January to April 2015 and the Scale-N baseline study from July to August 2016. Data were collected from mothers or caregivers from eight different rural villages, four of them in each of the two districts Chamwino and Kilosa of the Dodoma and Morogoro region, respectively. The study population had a total of 1010 women, 344 from the Trans-SEC and 666 from the Scale-N study.

Socio-demographic, anthropometric and nutritional data were collected from all 1010 women. Body Mass Index (BMI) was calculated and classified according to the WHO categories (WHO 1995). The Scale-N study included additional data on serum vitamin A (retinol binding protein), iron status (serum ferritin, soluble transferrin receptor) and serum zinc, measured at the VitMin Lab (Dr. J.G. Erhardt, Willstaett, Germany) using ELISA and photometric methods as previously described (Erhardt et al. 2004; Makino et al. 1982). Haemoglobin concentrations lower than 120 g/L were set as cut-off values for anaemia (WHO 2001). Iron deficiency was defined as ferritin <12 µg/L and/or soluble transferrin receptor (sTfR) >8.5 mg/L (Cook et al. 1992), while total body iron stores (IST) were calculated using SF and sTfR in an equation processed by Cook et al.: body iron (mg/kg BW) = - $[\log_{10}(sTfR \times 1000/SF) - 2.8229]/0.1207$ (Cook et al. 1992, 2003). Retinol <1.05 µmol/L was considered indicative of low vitamin A status (WHO 2001), while serum zinc <0.66 mg/L was used as cut-off for low/ deficient zinc status (Hotz et al. 2003).

Dietary assessment

Field surveys with questionnaires on maternal characteristics and dietary intake, including Dietary Diversity scores, FFQ and 24-h-recalls, were conducted. The 24-h-recalls were used for the in-depth evaluation of nutrient intake calculation by the CIMI algorithm, in comparison to NutriSurvey. Each woman was asked to recall all foods she had consumed during the day before the interview. Participants were requested to give quantities and ingredients of all foods consumed, during the whole day (Eleraky et al. 2022). Serving sizes were estimated and quantified using standardised plates, spoons and cups.

Nutrient intake calculation with the reference software NutriSurvey

NutriSurvey is a program that analyses the macroand micronutrient content of individually consumed foods. When NutriSurvey is used, it is necessary to enter each individual food and the exact amount consumed in grams. Nutrient compositions of local foods and dishes were added from the 'Tanzania Food Composition Table' (Lukmanji et al. 2008). The daily recommended nutrient intake values (RNI) for all micronutrients from the report 'Human Vitamin and Mineral Requirements' of a joint FAO/WHO expert consultation in Bangkok, were applied to assess the prevalence of insufficient micronutrient intake (FAO 2001; WHO 2001). Due to the predominantly plant-based diet in the study area, RNI for low zinc absorption rates and 10% iron bioavailability were applied (De Carli et al. 2018; WHO 2001).

When foods were added from the 'Tanzania Food Composition Table', in which vitamin A contents were provided as retinol activity equivalents, they were converted to retinol equivalents (RE) (Lukmanji et al. 2008). In addition, micronutrient contents (RE, Fe, Zn, Ca, Mg) for the local leafy vegetables Mugunda (Corchorus trilocularis) in Kilosa and Ilende (Ceratotheca sesamoides) in Chamwino were analysed by the Scale-N team in the laboratories of the University of Hohenheim (Gowele et al. 2017).

Identification of food groups for CIMI

Contrary to NutriSurvey, CIMI evaluates the 24-h-recalls by grouping the individually consumed foods into food groups to reduce data entry. Consumed foods in the two districts were split into 23 different food groups according to the FFQ data of the Trans-SEC sub-sample (FAO 2011). Composition of the food groups and contribution of individual foods to the different groups are shown in (Table 1).

Food items with high consumption levels, such as the main staple foods, usually form a separate group in contrast to other, less frequently consumed foods.

Table 1. Description of the food groups in CIMI-algorithm.

Due to high preference for maize and bulrush millet in the study population, both food items were separated from other foods within the category "other cereals". Furthermore, food items with high contents of particular nutrients were separated (e.g. vegetables were categorised into provitamin A-rich vegetables and other vegetables). The nutrient profiles of each food group were determined by calculating the average amount of nutrients with regard to the proportionate consumption of the food items in the target region that are aggregated in a specific food group. This was achieved using the following formula,

$$M_x = \frac{\sum_{k=1}^n c_k \cdot n_{xk}}{\sum_{k=1}^n c_k}$$

where M_x is the average content per gram of a nutrient x of a food group; c_k is the amount of an individual food item (grams) in the food group consumed, and n_{xk} is the content of nutrient x per gram of the individual food item k.

Statistics

All statistical analyses were conducted with SPSS (Version 27). P values <0.05 were considered as statistically significant. Social demographic data and blood concentrations (haemoglobin, retinol, zinc and iron

	Food Group	Contributing food items (% contents of each food)
1.	Maize	Maize flour dry (100%)
2.	Bulrush millet	Bulrush millet dry (100%)
3.	Sorghum	Sorghum flour dry (100%)
4.	Fried Bakery	Chapatti, African Donuts, Bhajia -> all three food items have an average contribution of 33.33%
5.	Other cereals	Rice (60%), finger millet (20%), wheat (20%)
6.	Plantain, white roots and tubers	Cooked bananas (30%), cassava (25%), white sweet potatoes (25%), potatoes (20%)
7.	Nuts	Groundnuts (90%), cashew nuts (10%)
8.	Pulses	Beans (kidney), cowpeas, pigeon peas, green grams, chicken peas, soya beans, Bambara nuts, lablab beans (hyacinths beans) -> all eight food items have an average contribution of 12.5%
9.	Vitamin A rich vegetables and yellow tubers	Yellow sweet potatoes (60%), pumpkin (25%), carrots (15%)
10.	Green leafy vegetables	Ilende [Ceratotheca sesamoides] (25%), cowpea leaves (20%), Mugunda [Corchorus trilocularis] (10%), amaranth (10%), sweet potato leaves (10%), Chinese cabbage (10%), cassava leaves (5%), spinach (5%), pumpkin leaves (5%)
11.	Other vegetables	Tomatoes (30%), onions (30%), African eggplant (30%), eggplant (5%), green pepper (5%)
12.	Meat	Beef (50%), chicken (15%), goat (15%), lamb (15%), pork (5%)
13.	Organ meat	Beef tripe (60%), beef liver (40%),
14.	Eggs	Chicken egg boiled (50%), chicken egg fried (50%)
15.	Milk and milk products	Yogurt (55%), cow's whole fat milk (40%), milk powder (5%)
16.	Fish (all types)	Sardines (40%), fish relish (30%), fried fish (20%), smoked fish (10%),
17.	Vitamin A rich fruits-A	Mangoes (50%), papaya (50%)
18.	Other fruits	Oranges (30%), watermelon (30%), ripe bananas (30%), pineapple (5%), avocado (5%)
19.	Red palm oil	Palm oil (100%)
20.	Other oils and fats	Coconut cream (60%), vegetable oil (40%)
21.	Added sugar/honey	Sugar (90%), honey (10%)
22.	Soft drinks	Carbonated drinks, Coca-Cola, fruit flavored drinks -> all three food items have an average contribution of 33.33%
23.	Beer/ local brews	Beer local grain (90%), beer commercial (5%), beer local-nonspecific (5%)

CIMI: Calculator of Inadequate Micronutrient Intake.

status) of the Scale-N women were presented by median (interquartile range), range (minimum – maximum), and percentage (number) for categorical and binary data; differences between districts were assessed using Mann-Whitney-U and Chi-Square tests, as appropriate.

The dietary intakes of macro- and micronutrients are given as means, medians and interquartile ranges; the comparisons between CIMI and NutriSurvey were assessed by differences in means, medians, bivariate correlation (Pearson) and Bland-Altman analysis (Bland-Altman plots).

Results

Socio-demographic characteristics for Trans-SEC and Scale-N women

A total of 1010 women of the Trans-SEC (N=334) and Scale-N (N=666) surveys, 479 from Chamwino and 531 from Kilosa district, were enrolled for the assessment of their nutritional status. Median age of all participating women was 37 years with a range between 19 and 85 years (Table 2). The median weight and BMI were 53.4 kg and 22.3 kg/m², respectively; both were higher in Kilosa compared to Chamwino. Women in Chamwino were significantly taller, but had a significantly lower BMI than those from Kilosa district. The overall higher proportion of women with overweight compared to underweight (27% vs. 6.8%) was observed, which was in particular evident in Kilosa (32% vs. 5.5%).

Dietary intake of macro- and micronutrients assessed by CIMI-algorithm vs. NutriSurvey

The differences between the two methods were marginal for most of the nutrients: the difference was

 Table 2.
 Socio-demographic characteristics of women who participated in Scale-N and Trans-SEC surveys.

	All	Chamwino	Kilosa	р
Women, N	1010	479	531	
Age [years] ^a	37 (30, 45) 19– 85	38 (30, 47) 20 -85	36 (30, 44) 19–81	0.029
Weight [kg]ª	53.4 (48.0, 60.5) 30.5– 112.4	52.9 (48.0, 59.9) 30.5 - 99.0	53.8 (47.9, 61.6) 34.4– 112.4	0.221
Height [cm]ª	154 (151, 158) 131–188	155 (152, 159) 136– 188	154 (149, 157) 131– 169	<0.001
BMI [kg/m²]ª	22.3 (20.5, 25.2) 14.4–47.4	22.1 (20.2, 24.4) 14.4 - 40,0	22.7 (20.7, 26.1) 14.8- 47.4	<0.001
BMI <18.5, % (n) ^b	6.8 (69)	8.4 (40)	5.5 (29)	0.001
18.5 < BMI < 25, % (n) ^b	66 (667)	70 (334)	63 (333)	
BMI ≥25, % (n) ^b	27 (274)	22 (105)	32 (169)	

Data are median (interquartile range), minimum-maximum^a and percentage (number)^b.

p-values: Mann-Whitney U or Chi Square test as appropriate.

<15% of the reference method's median for energy (1.8%), protein (4.0%), carbohydrates (2.2%), retinol equivalents (1.2%), folic acid (14,4%), vitamin B1 (10.0%), B2 (0.0%), B6 (0.0%), calcium (11.1%), iron (3.6%), and zinc (11.1%). However, calculation of fat and vitamin C by CIMI showed differences in median intake of 24.6% and 70.0%, respectively. Estimation of median nutrient intake was higher in CIMI for energy (26.8 kcal), fat (5.9 g), carbohydrates (5.4 g), retinol equivalents (5.1 µg), vitamin B1 (0.1 mg), vitamin C (13.9 mg) and iron (0.7 mg), whereas the calculated intake of protein (-1.4 g), folic acid $(-46.5 \mu \text{ g})$, calcium (-43.8 mg) and zinc (-0.9 mg) was lower. Regarding vitamin B12, both methods identified a very low average intake (CIMI 0.7 µg, NutriSurvey 1.0 µg), as 70% of all women did not regularly consume animal products (Table 3).

Bland-Altman Plots for energy, protein, iron and zinc were generated according to Giavarina et al. (Figure 1) (Giavarina 2015): The CIMI-algorithm and NutriSurvey provided comparable results, particularly at intakes below the recommended amounts. Overall, the calculations showed a negligible dispersion and were mostly within the defined confidence intervals (± 1.96 SD), and thus did not contribute to a significantly different micronutrient intake assessment between the two methods.

Haemoglobin and serum micronutrients of the Scale-N women

Haemoglobin, iron status and serum retinol and zinc concentrations of women who participated in the Scale-N study are summarised in Table 4. There was an overall high prevalence of anaemia (28%), iron deficiency (22%) and low serum zinc concentrations (22%) in the study population. In Kilosa, a significantly higher prevalence of anaemia and iron deficiency was measured compared to Chamwino district, but a lower prevalence of low serum zinc.

Association of micronutrient intake below the RNI with micronutrient deficiencies

The evaluation of CIMI and NutriSurvey with regard to the intake of vitamin A, iron and zinc below the recommended nutrient intake (RNI), and the subsequent diagnosis of anaemia and further biomarkers of micronutrient deficiencies revealed congruent results for both methods. An intake of vitamin A and iron below the RNI was associated with a significantly higher prevalence of anaemia and iron deficiency in

			CIMI			NI truit C					
	N					Madian	lop		CIMI - NUL		D D
F	N	Mean	Median	IQR	Mean	Median		Mean	Median	IQR	Pearson R
Energy [kcal]	1010	1031	1574	1025, 2111	1565	1517	998, 1994	65.3	26.8	-49.1, 156.8	0.931
Kilosa	479 531	1230 1992	1904	727, 1649 1480, 2437	1230 1868	1795	704, 1596 1415, 2288	-0.4 124.5	9.8 92.3	-40.3, 56.5 -68.0, 296.8	0.936
Protein [g]	1010	40.1	34.5	21.7, 52.5	41.5	34.9	21.3, 54.3	-1.4	-1.4	-4.8, 3.6	0.898
Chamwino Kilosa	479 531	27.5 51.5	23.9 45.4	14.8, 34.5 32.9, 64.7	29.2 52.5	23.3 47.7	15.0, 35.5 33.3, 65.0	-1.8 -1.0	-1.5 -0.7	-3.6, 0.4 -7.8, 7.0	0.887 0.876
Fat [g]	1010	44.4	32.5	16.4, 62.7	30.2	24.0	13.4, 39.6	14.1	5.9	0.5, 23.0	0.778
Chamwino Kilosa	479 531	24.4 62.3	18.0 55.9	9.5, 30.5 33.1, 81.1	21.6 38.0	15.8 33.4	9.2, 27.2 21.7, 47.2	2.8 24.3	1.6 18.0	-1.3, 5.6 5, 38.5	0.869 0.707
Carbohydrates [g]	1010	257.6	247.1	149.3, 342.8	261.3	241.9	147.7, 353.9	-3.6	5.4	-24.2, 17.1	0.929
Chamwino Kilosa	479 531	178.2 329.3	161.2 316.8	99.8, 231.6 245.7, 404.5	173.8 340.2	149.1 325.5	94.7, 229.3 251.5, 417.6	4.4 -10.9	10.5 -12.8	1.6, 18.1 –40.5, 16.6	0.891 0.899
RE [µg]	1010	515	384	115, 717	685	422	135, 904	-170	5.1	-342.1, 89.1	0.512
Chamwino Kilosa	479 531	639 402	515 195	319, 782 70, 573	894 496	663 200	343, 1196 47, 562	-254 -94.1	-7.5 9.1	–503.1, 90.5 –100.9, 83.7	0.492 0.480
Vitamin B1 [mg]	1010	1.2	1.2	0.8, 1.6	1.1	1.0	0.6, 1.4	0.1	0.1	0, 0.3	0.832
Chamwino Kilosa	479 531	1.1 1.3	1.0 1.2	0.8, 1.5 0.9, 1.6	1.0 1.1	0.9 1.1	0.5, 1.3 0.7, 1.5	0.1 0.2	0.1 0.2	0, 0.3 0, 0.3	0.853 0.808
Vitamin B2 [mg]	1010	1.1	1.0	0.6, 1.4	1.1	1.0	0.6, 1.5	0	0	-0.2, 0.1	0.768
Chamwino Kilosa	479 531	1.1 1.0	1.0 0.9	0.7, 1.5 0.6, 1.4	1.2 1.1	1.0 0.9	0.7, 1.5 0.6, 1.5	-0.1 0	0 0	-0.1, 0.1 -0.3, 0.2	0.738 0.805
Vitamin B6 [mg]	1010	1.5	1.4	1.0, 2.0	1.5	1.4	0.9, 1.9	0	0	-0.2, 0.3	0.810
Chamwino Kilosa	479 531	1.4 1.7	1.2 1.6	0.9, 1.8 1.1, 2.0	1.4 1.7	1.1 1.5	0.8, 1.7 1.1, 2.1	0 0	0.1 0	-0.2, 0.3 -0.1, 0.3	0.766 0.844
Folic acid [µg]	1010	289	255	176, 379	373	324	215, 481	-84.8	-46.5	-132.5, -7.1	0.773
Chamwino Kilosa	479 531	251 322	217 303	159, 323 202, 416	341 402	284 356	197, 449 244, 534	-90.1 -80.0	-50.5 -45.1	–119.8, –17.4 –143.0, 11.7	0.778 0.775
Vitamin B12 [µg]	1010	0.7	0	0, 0	1.0	0	0, 0.4	-2.7	0	0, 0	0.715
Chamwino Kilosa	479 531	0.3 1.1	0 0	0, 0 0, 1.6	0.3 1.5	0 0	0, 0 0, 1.8	0 0.5	0 0	0,0 -0.3, 0	0.705 0.719
Vitamin C [mg]	1010	45.7	33.1	17.8, 61.5	39.5	19.9	4.2, 52.1	6.1	13.9	-4.4, 29.8	0.534
Chamwino Kilosa	479 531	32.0 58.0	27.1 47.1	15.3, 42.4 26.0, 78.1	33.4 45.1	9.2 29.9	1.5, 44.6 11.9, 55.9	-1.4 12.9	12.4 16.4	–16.9, 26.9 –1.5, 34.7	0.318 0.685
Calcium [mg]	1010	434	369	237, 542	570	394	196, 749	-136	-43.8	-240.2, 113.4	0.534
Chamwino Kilosa	479 531	517 360	459 300	309, 622 188, 461	740 417	608 221	380, 945 135, 439	-222.9 -57.5	-163.7 55.1	-350.2, -22.8 -83.6, 207.0	0.771 0.212
lron [mg]	1010	25.2	21.8	14.1, 32.7	23.7	19.2	12.6, 29.0	1.5	0.7	-2.0, 5.8	0.775
Chamwino Kilosa	479 531	31.9 19.1	29.0 16.7	20.0, 39.9 10.9, 24.3	31.9 16.3	28.2 14.2	19.5, 43.0 10.2, 19.8	0 2.8	0 2.0	-2.6, 3.8 -1.5, 7.7	0.872 0.428
Zinc [mg]	1010	9.1	7.1	5.3, 9.9	7.8	8.1	5.9, 11.3	-1.2	-0.9	-2.1, 0	0.838
Chamwino Kilosa	479 531	8.4 7.3	7.6 6.9	5.3, 10.9 5.2, 9.1	9.5 8.7	8.2 8.0	5.8, 12.2 6.0, 10.7	-1.1 -1.4	-0.9 -1.0	-1.8, -0.3 -2.5, 0.2	0.895 0.745

Table 3. Comparison of means and medians of energy and nutrient intake calculated by CIMI and NutriSurvey in Chamwino and Kilosa district.

N: Number; IQR: Interquartile range; RE: retinol equivalents; Pearson correlation: all p-values were < 0.001.

both nutrient calculation methods. However, an intake of zinc below RNI was not directly related to lower serum zinc concentrations. The CIMI-algorithm identified slightly more women with a nutrient intake of vitamin A and zinc below the RNI who had anaemia, low serum retinol, or low serum zinc compared to the NutriSurvey software (Table 5).

Discussion

The overall aim of this study was to evaluate the accuracy of the new food group-based nutrient calculation algorithm CIMI compared to the reference software NutriSurvey with regards to the calculation of macro- and micronutrient intake. Data on dietary intakes (24-h-recalls) of 1010 women from the semi-arid rural Chamwino district and sub-humid rural Kilosa district were assessed, with a special focus on micronutrient intake below the recommended amount (Agency US D I Development E Limited TC Health R Survival C 2008).

The accuracy of CIMI was high for energy, protein, carbohydrates and most micronutrients. On the one hand, simplification of nutrient intake assessment by



Figure 1. Scatter plots (upper row) and Bland-Altman plots (lower row) of energy (A,B), protein (C,D), iron (E,F) and zinc (G,H) intake calculated by CIMI and NutriSurvey (NS). Scatter plots include predictive equations, vertical lines in the Bland-Altman plots represent the recommended nutrient intake values (RNI).

Table 4. Haemoglobin, iron status (iron stores, serum ferritin and soluble transferrin receptor) and serum retinol and zinc of the women who participated in the Scale-N survey.

	All	Chamwino	Kilosa	p
Women, N	666	333	333	
Hb [g/L] ^a	127 (118, 135)	131 (123, 139)	124 (113, 131)	< 0.001
Hb <120g/L, % (<i>n</i>) ^b	28 (187)	17 (56)	39 (131)	< 0.001
Retinol [µmol/L]ª	1.3 (1.6, 2.0)	1.7 (1.3, 2.0)	1.3 (1.1, 1.6)	< 0.001
Retinol <1.05, % (n) ^b	7 (48)	6 (21)	8 (27)	0.369
Iron stores [mg/kg BW] ^a	5.7 (3.9, 7.8)	6.3 (4.4, 8.2)	5.4 (3.0, 7.1)	< 0.001
Iron stores <0, % (n) ^b	7 (48)	5 (18)	9 (30)	0.072
SF <12 or sTfR >8.5 ^b	22 (147)	16 (54)	28 (93)	< 0.001
Zinc [mg/L] ^a	0.75 (0.68, 0.84)	0.73 (0.67, 0.82)	0.78 (0.70, 0.85)	0.001
Zinc <0.66 mg/L, % (<i>n</i>) ^b	20 (132)	25 (82)	15 (50)	0.002

Hb: Haemoglobin; SF: serum ferritin; sTfR: soluble transferrin receptor.

Data are median (interquartile range)^a and percentage (number)^b.

P-values: Mann-Whitney U and Chi Square test, as appropriate.

Anaemia is defined as Hb <120 g/L, low retinol as retinol <1.05 μmol/L, negative iron stores as iron stores <0 mg/kg BW, iron deficiency as SF <12 μg/L and/or sTfR >8.5 mg/L and low zinc status as Zinc <0.66 mg/L.

categorising single food items into specific food groups reduces data input, but on the other hand, the risk of deviation in nutrient calculation increases. This was more pronounced for nutrients with a high variation (e.g. vitamin A, vitamin C, folic acid and calcium) in foods belonging to one food group (e.g. 'green leafy vegetables', 'fish', 'pulses' or 'others fruits'), which showed less accuracy.

This phenomenon was also described by Jati et al. who investigated the food group-based approach in Indonesia (Jati et al. 2014) and by Bosha et al. in Ethiopia (Bosha et al. 2019b). In the present study, the results of CIMI-algorithm and NutriSurvey for energy, macronutrients and most of the micronutrients, were comparable with regard to the average intake. The lower correlation coefficient of vitamin A (0.512), 0.492 in Chamwino and 0.480 in Kilosa district, was due to the food groups' concept of CIMI. The food group 'green leafy vegetables' in CIMI calculates an average of all green leafy vegetables included in this food group e.g. Mugunda/Ilende (Corchorus trilocularis and Ceratotheca sesamoides), Cowpea leaves and sweet potato leaves. While the leafy vegetables Mugunda/Ilende or cowpea leaves, for instance, are very rich in provitamin A, other green leafy vegetables in the same group, such as cassava leaves or Chinese cabbage, do not contain high amounts of provitamin A. Women from Kilosa reported a high consumption of cowpea leaves, while women from Chamwino had a high intake of Ilende (Ceratotheca sesamoides). The predetermined average of the provitamin A content in the 'green leafy vegetables' group leads to an

Table 5. Prevalence o	f anaemia and	vitamin A, iro	n and zinc	micronutrient	deficiencies	in the	Scale-N	women	depending	on
nutrient intake (vs. <	and≥RNI) calc	ulated by CIMI-	algorithm	and NutriSurve	ey.					

	RE inta	ake < RNI	RE inta	ake≥RNI
Women % (<i>n</i>)	CIMI	NS	CIMI	NS
	n=412	n = 357	n = 254	n = 309
Anemia	31(126)**	29 (105)*	19 (47)	22 (68)
Retinol <1.05 μmol/L	8 (34)	7 (26)	6 (14)	7 (22)
SF < 12 and/or sTfR > 8.5	26 (106)*	26 (93)*	16 (41)	18 (54)
Zinc <0.66 mg/L	20 (81)	19 (68)	20 (51)	21 (64)
	iron in	take < RNI	iron in	take ≥ RNI
	CIMI	NS	CIMI	NS
	n = 493	n=522	n = 173	n = 144
Anemia	29 (145)**	29 (150)*	16 (28)	16 (23)
Retinol <1.05 μmol/L	8 (40)	8 (39)	5 (8)	6 (9)
SF < 12 and/or sTfR > 8.5	25 (121)*	24 (124)*	15 (126)	16 (23)
Zinc <0.66 mg/L	19 (91)	19 (101)	24 (41)	22 (31)
	zinc in	take < RNI	zinc in	take≥RNI
	CIMI	NS	CIMI	NS
	n = 554	n = 499	n = 112	n = 167
Anemia	27 (151)	26 (129)	20 (22)	26 (44)
Retinol <1.05 μmol/L	8 (43)	7 (34)	5 (5)	8 (14)
SF < 12 and/or sTfR > 8.5	23 (127)	24 (118)	18 (20)	17 (29)
Zinc <0.66 mg/L	20 (113)	21 (104)	17 (19)	17 (28)

All data are presented in percentages and numbers in brackets. P-values (Chi-square test): *<0.05, **<0.001.

Anaemia is defined as Hb <120 g/L, low retinol as retinol <1.05 µmol/L, iron deficiency as serum ferritin (SF) <12 µg/L and/or soluble transferrin receptor (sTfR) >8.5 mg/L and low zinc status as zinc <0.66 mg/L.

RNI: recommended nutrient intake: RE 500 µg/d, iron 29 mg/d (10% bioavailability), zinc 9.8 mg/d (low bioavailability) (De Carli et al. 2018).

underestimation of vitamin A intake compared to NutriSurvey, which only accounts for the actual consumed leafy vegetables. Similarly, the participants of Kilosa stated consuming ripe bananas and watermelon, but they hardly reported eating oranges. Nevertheless, the group 'fruits' in CIMI calculated an average of all fruits included in this food group. Oranges are vitamin C-rich fruits and contribute to 30% of the group's nutrient composition, which led to an overestimation of vitamin C intake by the CIMI-algorithm in regions where no, or only few oranges were consumed.

Although the CIMI results verged on the calculation of NutriSurvey for most of the micronutrients, some differences in the correlation coefficients occurred between both districts, particularly concerning vitamin C, calcium and iron. With regards to calcium, there was an underestimation of consumption in CIMI results compared to NutriSurvey, especially for the women from Chamwino, who mainly consumed the calcium-rich leafy vegetable Ilende (Ceratotheca sesamoides). This underestimation, similar in the case of vitamin A, is due to the food groups concept of CIMI, which calculates the average of the calcium content of the 'green leafy vegetables group', which is clearly lower than the calcium content of Ilende. Ilende was the primarily and almost exclusively consumed green leafy vegetable in Chamwino, but rarely consumed in Kilosa, where multiple other vegetables with varying calcium and iron contents were consumed. The higher correlations in Chamwino with regard to calcium and iron can be explained by the little varied dietary habits of the women from this region, who consumed less food items than the more food diverse women from Kilosa. The low correlation regarding the calcium intake in Kilosa was primarily due to the consumption of sardines (small dried fish with eatable fish bone), which include extremely high amounts of calcium. Consumers of sardines were mainly located in Kilosa district. This led to statistical outliers showing very high calcium intake values in NutriSurvey and lower ones in CIMI, resulting in a weaker correlation for calcium. Nevertheless, the food group-based approach of CIMI is suitable to estimate nutrient intake at the population level.

In order to increase the accuracy on the individual level, further fragmentation of the food groups or optional deselection of not consumed foods within a food group would increase the precision of nutrient calculation. Such a CIMI already exists for Ethiopia, showing high validity, requires less time and is well accepted, by interviewers as well as for interviewed participants (Bosha et al. 2019a, 2019b).

One of the advantages of CIMI is the fast and simple data input, since only the consumed amounts of the different food groups have to be entered. Results of nutrition analysis are available immediately after the interview, which enables the interviewer to inform participants about possible individual nutrient deficits and available food sources to close the gap (Cook et al. 1992). In addition, individual feedback to the participants possibly increases the willingness to participate in such surveys and motivates the participants to provide correct answers.

Our study had some limitations: While the dietary patterns of the Tanzanian population vary greatly, only two agro-ecological zones in Tanzania where included. There are different main staples: maize is preferred in the West and South, millet in the middle and rice in the East. In addition, dietary assessment was only done in rural areas, where the meals are mainly prepared at home. Assessing nutrient intake in urban settings is much more challenging, due to the possibility to buy convenience food or eat out-of-home.

The nutrient intake calculation in our study was based on only one 24-h-recall, which limits the quality of the results. According to Lombard et al. one 24-h-recall analysis is not sufficient to provide valid dietary intake results and more than three 24h-recalls are required to obtain valid data (Lombard et al. 2015). Furthermore, retrospective categorisation of 24-h-recall data into food groups might confound the results. Nevertheless, the CIMI-study in Ethiopia clearly demonstrated that the direct data input in CIMI during the interview, generated comparable results to NutriSurvey and showed a time saving for the interview of 25%, which did not take into account the time needed for classical 24-h-recall data analysis by nutrition software (Bosha et al. 2019a).

As our findings have shown, CIMI calculation is prone to over- or underestimation of micronutrient intake due to fixed nutrient profiles for entire food groups. Therefore, identification of individual nutritional gaps by the CIMI-algorithm is less specific. Integration of an option enabling deselection of not-consumed foods in a food group might help to overcome this limitation. However, for rapid population-level assessment of nutrient intakes, the simple food group-based CIMI algorithm is appropriate and a useful tool for generating data needed to plan, implement, and measure the impact of intervention programs.

In conclusion, CIMI is a suitable tool for estimating nutrient intakes in rural Tanzania. Its application provides detailed information on energy, macro- and micronutrient intakes without requiring additional time for data analysis. Therefore, CIMI can be used in place of nutrient diversity assessment, which also does not provide quantitative data on intake of individual nutrients. Additional enhancements to CIMI, such as automatic calculation of percent compliance with nutrient intake recommendations and the ability for respondents to deselect foods within a food group, will further enhance usability.

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Authors' contribution

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