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The ability of anthropometric indices in detecting type 2 diabetes mellitus in Sudanese adults: a cross-sectional study

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Diabetes mellitus (DM) is a growing global health problem. We aimed to evaluate the ability of body mass index (BMI), waist circumference (WC), and waist-height ratio (WHtR) to detect type 2 diabetes mellitus (T2DM) among adults in Northern Sudan. A multi-stage sampling survey was conducted in Northern Sudan. The information was obtained on the sociodemographic and anthropometric indices (BMI, WC, and WHtR). A receiver operating characteristic (ROC) curve and a multivariate binary analysis were performed. Among the 396 adults included in the study, 201 (50.8%) were male, and 195 (49.2%) were female. The median (interquartile, IQR) for age, BMI, WC, and WHtR were 45.0 (33.3–55.8) years, 26.5(22.5–30.5) kg/m², 84.0(73.8–95.0) cm, and 0.46 (0.40–0.53), respectively. The median values of BMI, WHtR, and WC were higher in females than in males. Of the total, 107 (27.0%) adults had T2DM. The median (IQR) of BMI, WC, and WHtR were significantly higher in adults with T2DM than in those without T2DM. BMI (AUC = 0.55, cutoff = 25.17 kg/m², sensitivity = 0.68, specificity = 0.48, PPV = 32.6, NPV = 80.2), WHtR (AUC = 0.62, cutoff = 0.44, sensitivity = 0.74, specificity = 0.40, PPV = 34.4, NPV = 84.8), and WC (AUC = 0.63, cutoff = 79.28 cm, sensitivity = 0.78, specificity = 0.47, PPV = 35.3, NPV = 85.1), had poor predictability in detecting T2DM in adults. In multivariate analysis, being female (adjusted odds ratio [AOR] 1.66, 95% CI 1.01–2.73), increased age (AOR 1.04, 95% CI 1.03–1.06), and increased WC (AOR 1.02, 95% CI 1.01–1.04) were associated with T2DM. BMI and WHtR were not associated with T2DM. BMI, WC, and WHtR showed poor predictive value for detecting T2DM among adult Sudanese.

Keywords Diabetes mellitus, Age, Body mass index, Waist circumference, Risk factors, Sudan

Abbreviations

DM	Diabetes mellitus
AOR	Adjusted odds ratio
BMI	Body mass index
MUAC	Mid-upper arm circumference
WC	Waist circumference
WHtR	Waist-height ratio
WHR	Waist–hip ratio
CI	Confidence interval
SD	Standard deviation
IQR	Interquartile range
AUC	Area under the curve
ROC	Receiver-operating characteristic curve

Diabetes mellitus (DM) is one of the most common chronic diseases associated with significant morbidities, mortalities, and economic burdens^{1,2}. It has been estimated that around half a billion people are living with DM worldwide, and this figure is expected to increase to 12.2% (783.2 million) in 2045^{1,3}. T2DM is the most common type of diabetes, accounting for over 90% of all diabetes worldwide¹. A higher prevalence of DM was reported

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in high-income countries (11.1%) compared to low-income countries (5.5%), and in urban (12.1%) than rural areas (8.3%)³. Half of adults (20–79 years old) with DM were unaware of their diabetes status (44.7%; 239.7 million); almost 90% of people with undiagnosed DM live in low- and middle-income countries, in particular among the African population (53.6%)^{1,4,5}. Patients with undiagnosed DM had considerable complications related to DM, such as microvascular and macrovascular complications⁵. In North Africa, it has been reported that the prevalence ranged from 2.6% in rural Sudan to 20.0% in urban Egypt⁶. Likewise, a higher prevalence of undiagnosed DM (18–75%) was reported in Northern Africa, with a higher prevalence of chronic complications related to T2DM⁶. On the other hand, maintaining reasonable glycaemic control is the cornerstone of preventing or modifying DM-related complications⁷. Thus, screening for T2DM, especially among higher-risk groups, is of paramount importance and may improve T2DM prognosis and minimize its complications. It is worth mentioning that some microvascular complications are two-to-three-fold higher in youth with T2 DM compared to those with type 1 DM of a similar age group⁸.

Body mass index (BMI) has been widely used as a convenient and straightforward measure of adiposity and obesity in many epidemiological studies, despite limitations in assessing body fat distribution or differentiating between body fat and lean mass⁹. There was substantial supporting evidence indicating that visceral adiposity, rather than overall body adiposity, is associated with metabolic complications¹⁰. Hence, anthropometric indices such as BMI, mid-upper arm circumference (MUAC), waist circumference (WC), waist-to-hip ratio (WHR), and waist-to-height ratio (WHtR) are widely advocated to assess visceral adiposity¹⁰. Visceral adiposity is associated with cardiometabolic syndrome: insulin resistance, hypertension, hyperlipidemia, and hyperuricemia¹¹. Weight, WC, and BMI are the most reliable indices for predicting T2DM in low- and middle-income countries¹². WC, WHR, and WHtR are used to assess body fat distribution and obesity, as they are simple, practical, and low-cost tools¹³. Several previous studies have demonstrated the significant utility of these valuable anthropometric indices for screening and diagnosing T2DM across different populations worldwide^{14–21}.

According to the IDF, Sudan is among the countries with a DM prevalence above 12%¹. This was supported by recently published studies indicating high prevalence in northern Sudan (19.1%)²² and eastern Sudan (20.8%)²³. Recent data demonstrated a considerable percentage of newly diagnosed T2DM (10.0–31.3%) and markedly higher prevalence of uncontrolled T2DM (80.0–84.3%)^{23,24}. Several studies from Sudan reported higher prevalences of overweight and obesity in the general population in northern Sudan (36.37% and 24.53%)²², (33.6% and 26.8%)²⁵, and in eastern Sudan (26.5% and 32.7%)²³. Likewise, a population-based survey conducted in Sudan showed a higher prevalence of overweight (39.9%) and obesity (24.5%) among individuals with T2DM²⁵. A higher prevalence of obesity among patients with T2DM was reported in eastern Sudan (32.2%)²⁶. Among other countries, Sudan lacks accurate data and well-established, sophisticated health systems to assess and provide comprehensive management for this patient group⁶. Moreover, Sudan is among the countries with the least diabetes-related health expenditure (USD) per person with T2DM (less than 250 dollars per year)¹. Additionally, no data on the accuracy of anthropometric indices in screening for T2DM in Sudan have been published. Moreover, obesity may be influenced by ethnicity, genetic, and epigenetic factors, which may be reflected in anthropometric measurements across different populations²⁷. Hence, in clinical settings, to assess non-communicable diseases, the cutoff point might be adjusted to increase the screening test's sensitivity and enhance the ability to identify potentially significant health outcomes²⁸.

Moreover, some studies have highlighted Africa's unique racial composition, genetics, culture, socioeconomic factors, dietary patterns, geography, environment, and political structures, as well as the need for Africa-specific studies to address the risk factors associated with T2DM^{29,30}. Additionally, understanding the specific relationships between potentially modifiable risk factors, such as biological, behavioral, and psychosocial factors, and T2DM in African populations is of paramount value for effective primary prevention and earlier detection efforts to modify the prognosis^{29,30}. However, no prior studies have evaluated the diagnostic accuracy of anthropometric indices in identifying T2DM among Sudanese adults. Hence, to address all points mentioned above, we conducted this study to assess the ability of BMI, WC, and WHtR to detect T2DM among Sudanese adults in Wad Hamid locality, northern Sudan.

Materials and methods

Study design and population

The study was conducted in accordance with the Declaration of Helsinki and was approved by the Almatamah Health Authority, Sudan, under reference #9, 2021. All participants recruited for the study signed written informed consent forms after proper explanations regarding the study and its aims. The authors had a direct follow-up to ensure all necessary measures, including participants' privacy, confidentiality, and safety, were taken, such as excluding personal identifiers during data collection.

It was a cross-sectional study targeting Sudanese adults aged 18 and above from September 1 to December 31, 2022. The Strengthening the Reporting of Observational Studies in Epidemiology (STROBE) guidelines were strictly followed for study³¹. A written informed consent form was gathered from all participants, indicating their acceptance of participation in the study. A multi-stage sampling study was conducted in Wad Hamid city, a locality in the River Nile State in Northern Sudan. The residents of Wad Hamid are Arab tribes, and minorities are Nubian and African tribes, who have a long history of living along the Nile River in Northern Sudan. Wad Hamid locality is one of the four localities (the smallest administrative unit in Sudan) in the River Nile State. Initially, random sampling was performed, and the locality was divided into four sectors based on their population size. A lottery method was used to randomly select all adult participants in households in these four sectors. If no candidates were available in a selected household, or household members refused to participate or met any exclusion criteria, the next household was an alternative.

Inclusion criteria

All healthy adult Sudanese citizens, both male and female, residing in the study area and aged 18 years or above, agreed to participate with written informed consent. Moreover, all participants with T2DM on diet control or glucose-lowering drugs were included.

Exclusion criteria

Age below 18 years, individuals with chronic diseases (based on participants interview regarding diagnosis, specialty follow up, medical report and current medications) such as chronic kidney diseases and heart failure, critically ill patients with severe acute illness, those on hormonal or chronic medication (e.g. steroids), those who had any apparent congenital dysmorphism, and those with missing data and who refused to participate in the study. In addition to T1DM, pregnant women, patients with hemoglobinopathies, or current medications like steroids and hormonal contraceptives were excluded, too.

Participants

Each participant who signed the informed consent forms was interviewed using a comprehensive questionnaire, designed based on data collection tools from previous studies^{22,23,25,26}. The sociodemographic characteristics were obtained in the same questionnaire: age in years, sex, marital status, job, education, smoking habits, drug history, and family history of T2DM. The anthropometric indices: weight, height, WC, BMI, and WHtR) and sociodemographic characteristics were obtained in the same questionnaire: age in years and sex. Two senior-trained nurses were assigned to collect the anticoagulated venous blood sample (5 mL) and mix it gently before sending it immediately to measure HbA1c.

Diabetes mellitus

Adults were considered to have T2DM if they were already diagnosed before with T2DM by a physician, and were on diet control or glucose-lowering drugs during the time of the study, or if their HbA1c $\geq 6.5\%$ as per the guidelines and recommendations of the “International Diabetes Federation guideline for non-pregnant adults”³². Ethylenediaminetetraacetic acid (EDTA) was used as an anticoagulant for the collected venous blood sample (5 mL), which was then mixed gently and sent directly for laboratory investigations. HbA1c was tested using an Ichroma machine following the manufacturer’s instructions (Chuncheon-si, Gang-won-do, 24398, Republic of Korea). Calibration was stable for at least fifty days. The process was under the control of two primary investigators, and the central laboratory measures.

Four trained medical officers interviewed participants during the study period, and eight trained paramedical staff collected blood samples, measured participants’ anthropometric indices, and recorded these measurements on the questionnaire sheet. To improve the accuracy of the collected data, key quality control measures include using certified medical personnel, adopting standard operating procedures, ensuring equipment is robust and well-calibrated, and training all staff to meet specific measurement standards and inter-observer reliability targets, under the direct supervision of two primary investigators.

Anthropometric measurements

Generally, all measurements were checked twice to get the averages for the study. A third reading was performed when significant differences were observed between the first two readings across all measures.

Weight and height measurements

We ensured that all participants wore light clothing and no shoes while measuring weight and height. The well-calibrated scales of standard procedures were used to measure participants’ weights in kilograms, adjusted to zero before each measurement and to the nearest 100 g. The participants were asked to stand straight, with their backs against the wall and their feet together, allowing their heights to be measured to the nearest 0.1 cm.

BMI

BMI was calculated by dividing participants’ weight (kg) by their height (m²).

WC

The participants adopted a relaxed standing position and were directed to fold their arms. Seca 203 tape (made in Germany) was used for measuring WC. Then WC was measured in centimeters with a non-stretchable tape held at the level of the navel while the subject was standing without clothing at the end of the normal expiration at the midpoint between the superior border of the iliac crest and the lower margin of the least palpable chest rib³³.

WHtR

WHtR was calculated by dividing WC (in centimeters) by height (in centimeters).

Sample size

The sample size of 396 participants was calculated based on the assumption that 25.0% of adults would have T2DM, as reported previously³⁴, and that anthropometric measures (BMI and WC) would have a sensitivity of 90.0% to detect T2DM. It is expected to have 80% power and a difference of 5% ($\alpha = 0.05$)³⁵.

Statistical analysis

IBM Statistical Package for the Social Sciences® (SPSS®) for Windows, version 22.0 (SPSS Inc., New York, United States), was used to analyze the Data. Frequencies (%) were adopted to express the proportions. The Shapiro-

Variables	Total	Adults with type 2 diabetes mellitus (number = 107)	Adults without type 2 diabetes mellitus (number = 289)	P
Body mass index, kg/m ²	26.5(22.5–30.5)	27.1(23.5–30.8)	25.7(22.0–30.4)	< 0.001
Waist circumference, cm	84.0(73.8–95.0)	88.5 (80.0–98.5)	81.0(71.0–93.5)	< 0.001
Waist-to-height ratio	0.46 (0.40–0.53)	0.50(0.44–0.55)	0.45(0.38–0.52)	< 0.001

Table 1. Comparing the median (interquartile range) of the adults with and without diabetes mellitus in Northern Sudan, 2022. The Mann–Whitney U test was used.

Variable	Males	Females	P value
Body mass index, kg/m ²	24.9 (21.5–29.0)	27.4 (23.9–31.2)	< 0.001
Waist circumference, cm	82.0 (71.0–92.5)	86.0 (76.0–96.0)	0.014
Waist-to-height ratio	0.44 (0.38–0.52)	0.49 (0.42–0.54)	0.001

Table 2. Comparing the median (interquartile range) of the anthropometrics between males and females in Northern Sudan, 2023. The Mann–Whitney U test was used.

Variable	Glycated hemoglobin (HbA1c) level	
	R	P value
Body mass index, kg/m ²	0.088	0.079
Waist-to-height ratio	0.065	0.194
Waist circumference, cm	0.162	0.001

Table 3. Spearman correlation between anthropometric measures and HbA1c level in adults in Northern Sudan, 2023.

Wilk test was used to assess the normality of continuous data (age, height, weight, BMI, WC, WHTR, and HbA1c readings), which showed nonnormal distributions; therefore, they were reported as medians (interquartile ranges [IQR]). Medians were compared between the two groups using the Mann–Whitney U test. Correlations between anthropometric indices and HbA1c levels were assessed. A receiver operating characteristic (ROC) analysis was performed to determine the predictive power of various body composition indices for T2DM diagnosis among Sudanese adults. The area under the receiver operating characteristic (ROC) curve (AUC) was used to assess the predictive power of these anthropometric indices for detecting T2DM. The obtained AUCs were classified according to their power: 1 indicates a perfectly accurate test; > 0.9 is excellent, 0.8–0.9 is good, 0.7–0.8 is acceptable, < 0.7 is considered poor, and ≤ 0.5 suggests that the test has no discriminatory ability³⁶. Optimal anthropometric index cutoff values for predicting T2DM were chosen based on the most considerable value obtained from the Youden index ($J = SE + Sp - 1$)³⁷. Sensitivity, specificity, Youden index, positive and negative predictive values (PPV and NPV) were computed to assess the validity of the different anthropometric indices for predicting T2DM. The acceptable value of sensitivity and specificity of the screening tools adopted in this study was 70.0%³⁸. Multivariate binary analysis was performed with T2DM as the dependent variable and age, sex, and anthropometric measurements as independent variables. Multicollinearity was checked using the Variance Inflation Factor (VIF < 4), and none was detected. The goodness-of-fit test (Hosmer–Lemeshow Test) showed a good fit. A p-value of < 0.05 was used to indicate the significance level for all performed tests.

Results

General characteristics

Among the 396 adults included in the study, 201 (50.8%) were male and 195 (49.2%) were female. Of 396 adults, 267 (67.4%) had an education level beyond secondary, 182 (46.0%) were employed, and 297 (75.0%) were married. Eighty-two (20.7%) adults were smokers, and 146(36.9%) had a family history of DM. The median (IQR) for age, BMI, and WC were 45.0 (33.3–55.8) years, 26.5 (22.5–30.5) kg/m², and 84.0 (73.8–95.0) cm, respectively (Table 1).

The median values for BMI, WHTR, and WC were higher in females than in males (Table 2). Of the total adults, 107 (27.0%) were identified with T2DM. The median (IQR) of BMI, WC, and WHTR were significantly higher in adults with T2DM than in those without T2DM (Table 1). There was no correlation between BMI, WHTR, and HbA1c levels. There was a positive correlation between WC and HbA1c (Table 3).

For the total adults (both males and females), BMI (AUC = 0.55, cutoff = 25.17 kg/m², sensitivity = 0.68, specificity = 0.48, PPV = 32.6, NPV = 80.2), WHTR (AUC = 0.62, cutoff = 0.44, sensitivity = 0.74, specificity = 0.40, PPV = 34.4, NPV = 84.8), and WC (AUC = 0.63, cutoff = 79.28 cm, sensitivity = 0.78, specificity = 0.47, PPV = 35.3, NPV = 85.1) had poor predictability in detecting T2DM in adults (Table 4; Fig. 1).

Variable		The area under the curve	95.0% confidence interval	Cut-off point	Sensitivity	Specificity	Youden's index	Positive predictive value	Negative predictive value
Body mass index, kg/m ²	Both groups	0.55	0.49-0.61	25.17	0.68	0.48	0.16	32.6	80.2
	Males	0.59	0.50-0.68	23.0	0.78	0.41	0.18	30.9	81.7
	Females	0.50	0.41-0.59	25.3	0.74	0.39	0.14	33.9	77.9
Waist-to-height ratio	Both groups	0.62	0.59-0.75	0.44	0.74	0.40	0.14	34.4	84.8
	Males	0.67	0.59-0.74	0.44	0.67	0.44	0.11	33.9	87.0
	Females	0.56	0.48-0.65	0.45	0.78	0.39	0.16	34.8	81.7
Waist circumference, cm	Both groups	0.63	0.57-0.68	79.2	0.78	0.47	0.25	35.3	85.1
	Males	0.65	0.57-0.74	80.5	0.80	0.56	0.36	36.8	89.5
	Females	0.59	0.51-0.68	76.9	0.86	0.31	0.18	34.7	84.3

Table 4. Performance of anthropometric measures to detect type 2 diabetes mellitus among adults in Northern Sudan, 2023.

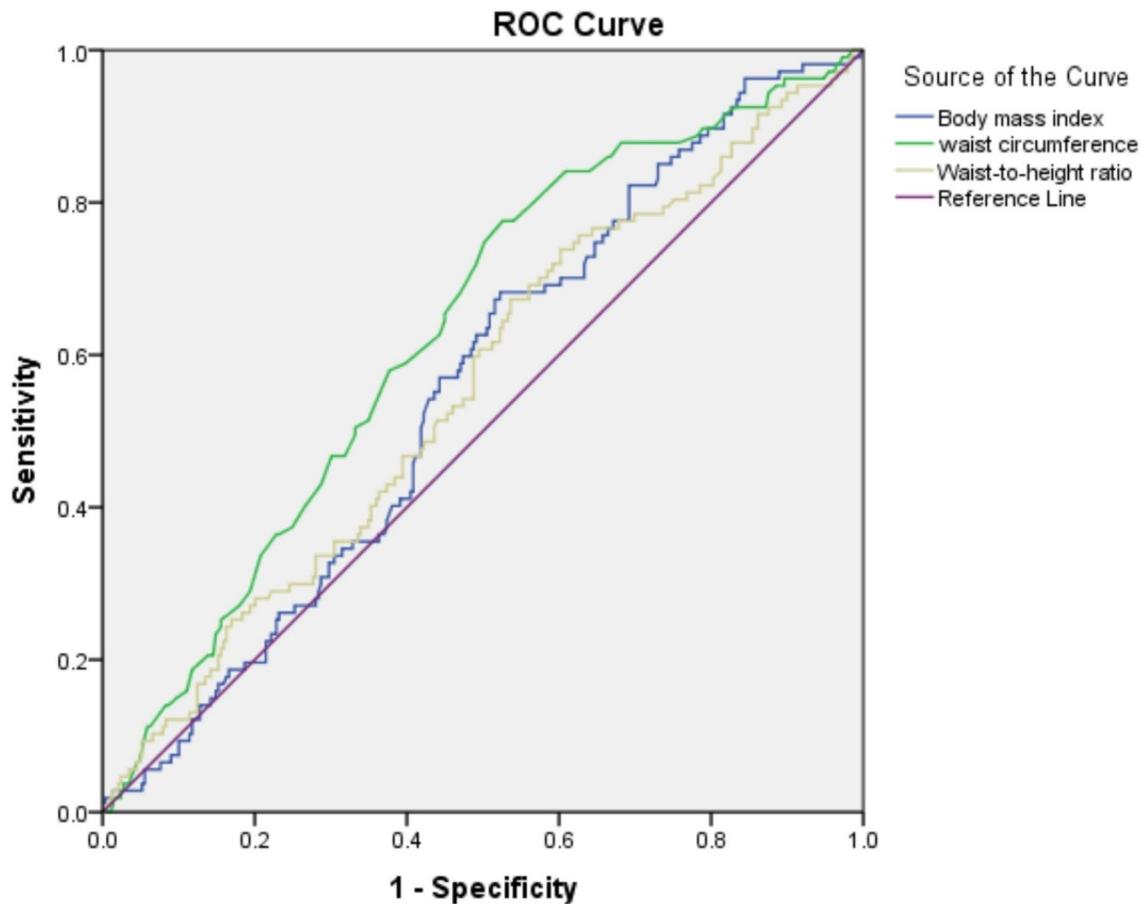


Fig. 1. Receiver operating characteristic (ROC) curves of BMI, WC, and WHtR for detecting Type 2 diabetes mellitus among adults in Northern Sudan.

When the results were dissected according to sex into males and females, for males, BMI (AUC=0.59, cutoff=23.0 kg/m², sensitivity=0.78, specificity=0.41, PPV=30.9, NPV=81.7) and WHtR (AUC=0.67, cutoff=0.44, sensitivity=0.67, specificity=0.44, PPV=33.9, NPV=87.0) had poor predictability in detecting T2DM. WC had a slightly higher AUC (AUC=0.65; cutoff=80.5 cm; sensitivity=0.80; specificity=0.56, PPV=36.8, NPV=89.5) (Table 4).

In females, BMI (AUC=0.50, cutoff=25.3 kg/m², sensitivity=0.74, specificity=0.39, PPV=33.9, NPV=77.9), WHtR (AUC=0.56, cutoff=0.45, sensitivity=0.78, specificity=0.39, PPV=34.8, NPV=81.7), and WC (AUC=0.59, cutoff=79.9 cm, sensitivity=0.86, specificity=0.31, PPV=34.7, NPV=84.3) had poor predictability in detecting T2DM in females' adults (Table 4).

Variable	Odds ratio	95% confidence interval	P value	Adjusted odds ratio	95% confidence interval	P value
Age, years	1.04	1.02–1.06	<0.001	1.04	1.03–1.06	<0.001
Sex, females vs. males	1.31	0.84–2.05	0.230	1.66	1.01–2.73	0.046
Body mass index, kg/m ²	1.03	0.99–1.06	0.137	0.97	0.92–1.02	0.289
Waist-to-height ratio	1.12	0.75–1.67	0.564	0.45	0.18–1.14	0.094
Waist circumference, cm	1.02	1.01–1.03	0.008	1.02	1.01–1.04	0.011

Table 5. Univariate and multivariate analyses of the factors associated with diabetes mellitus in adults in Northern Sudan, 2023.

The univariate analysis indicated that increased age and WC were associated with T2DM. In multivariate analysis, being female (AOR 1.66, 95.0% CI 1.01–2.73), increasing age (AOR 1.04, 95% CI 1.03–1.06), and increased WC (AOR 1.02, 95% CI 1.01–1.04) were associated with T2DM; BMI and WHtR were not associated with T2DM (Table 5).

Discussion

The current study showed that BMI, WC, and WHtR were significantly higher in females than in males. Over the quarter, (27.0%) of adults had T2DM, and females were at higher risk of having T2DM (AOR = 1.66). The higher prevalence of T2DM reported in our study concords with a global increase in T2DM, particularly in Africa¹, and is supported by recently published studies evaluating the prevalence of T2DM among Sudanese adults^{23,24}. Additionally, the increased life expectancy due to improved medication, improved awareness, lifestyle changes, and health facilities for early detection of T2DM might be contributing factors²³. In the current study, females were at a higher risk of T2DM. A higher prevalence of T2DM among females was reported in previous studies conducted in Sudan, which showed that overweight, obesity, and T2DM were more prevalent in Sudanese females than males^{23,26}. Rapid urbanization may enhance the prevalence of obesity in African countries³⁹. Likewise, another study documented the pronounced obesity among women in developing countries⁴⁰. Additionally, obese females tend to have lower rates of physical activity when compared to males⁴¹. Moreover, biological and social factors, levels of physical activity, sociocultural beliefs, and urbanization enhance the influence of gender differences^{40,42}.

Our study revealed that BMI, WC, and WHtR, and their associated cut-off points, were significantly higher in female participants than in male participants. This is similar to results reported in recent publications from different African countries: South Africa^{19–21}, Egypt¹⁸, and Ghana⁴³. The higher cut-off points of the anthropometric indices seen among females in this study may be explained by the higher prevalence of obesity among females compared to males documented in different regions of Sudan, Khartoum, Northern, and River Nile States. Obesity in women (43.6%) is higher than in men (21.8%)²⁵. In Eastern Sudan, women are 36.6%, men 21.6%²⁸, and from Khartoum, the capital of Sudan, to the border of Egypt in Halfa city, females than males (36.7%) than men (17.6%)⁴⁴. Not only Sudan but also adults from North Africa had a similar higher prevalence of overweight and obesity in females than in males⁴⁵. Rapid urbanization may contribute to a higher prevalence of obesity in African countries^{39,46}.

The current study showed that BMI, WC, and WHtR have limited discriminatory ability for detecting T2DM. A similar poor discriminatory range was reported in recently published clinical studies in African countries, such as Ethiopia, Cameroon, and Nigeria^{14,15,47}. On the other hand, some recently published studies in Africa documented a wide range of discriminatory ability for these robust anthropometric indices in detecting T2DM among the African population^{18,19,21,43}. The variation in the discriminatory ability of these indices may be explained by the fact that the distribution of fat tissue is the crucial factor in developing insulin resistance, independently of the degree of obesity, in particular, the ectopic fat deposition in the liver of obese patients is considered a significant marker of insulin resistance and glucose deregulation⁴⁸. Moreover, T2DM is a multifactorial disease caused by oligo- and polygenic genetic and non-genetic factors arising from an imbalance between energy intake and output, as well as other lifestyle-related factors such as diet, regular exercise, and social beliefs^{49,50}. This offers a better understanding of T2DM as a multifaceted disease, and its characteristics are essential for managing patients⁵⁰. Identifying the best anthropometric measurement for predicting T2DM remains contentious. Several previous studies have shown that BMI was the best predictor of T2DM risk among adults compared to WC and WHtR, e.g., in Nigeria¹⁵, Egypt¹⁸, South Africa^{19–21}, Jordan¹⁶, and China¹⁷. Analysis of biobank data from 450,000 British participants over 10 years revealed that BMI was the most significant predictor of T2DM among males, and WC was the most significant predictor among females⁴⁶. In contrast, some African studies identified WC, WHtR, and BMI as the best anthropometric indices for detecting T2DM, in order^{16,20,21,43,48}. Several studies worldwide reported no significant differences among the anthropometric indices (WC, WHtR, and BMI) and found them to be similar in predicting T2DM^{51–55}. The outcome of these studies comparing the predictive power of anthropometric indices for T2DM has yielded conflicting results; hence, using a combination of these measurements is a robust predictor and preferable to using them individually^{9,56,57}. Moreover, variations between the current study's findings and those of other studies might be attributable to many factors, including population distinctiveness, cultural dynamics, ethnic groups, physical activities, sampling techniques, data-collection methods, and differences in operational definitions⁴⁷. Additionally, IDF recommends the use of pre-specified cut-off points for BMI, WC, WHR, and WHtR to standardize comparisons within and between populations and ethnic groups, as the current cut-off points adopted by the World Health

Organization (WHO) and IDF slightly underestimate the screening of T2DM among other ethnicities, especially in men¹.

Our study demonstrated that anthropometric indices are poor screening tools for T2DM in resource-limited settings, particularly given the marked increase in T2DM prevalence in low- and middle-income countries. In fact, these indices remain practical screening tools in low-resource settings despite limited accuracy demonstrated in this study. The result obtained in our study might be explained by the fact that some participants were diagnosed and on lifestyle management or anti diabetic medications that influence obesity and ultimately affect the anthropometric indices^{58,59}. Additionally, T2DM is type 2 diabetes, which is multifaceted, 49 and genetic differences among racial/ethnic groups could also provide some explanation for the results obtained in this study⁶⁰. However, these anthropometric indices are simple, easily obtainable, low-cost, noninvasive measures of population screening and early detection of malnutrition. ^{61,62} Additionally, traditional measures like BMI, WC, and WHtR, as well as novel indices such as novel [a-body shape index (ABSI), abdominal volume index (AVI), body adiposity index (BAI), body roundness index (BRI), conicity index (CI), and Clinical Universidad de Navarra-Body Adiposity Estimator (CUN-BAE)] anthropometric indices, had similar performance in detecting T2DM^{20,63,64}. Moreover, the traditional anthropometric indices have a significant ability to predict the incidence of T2DM over the years⁶⁵ assess glycemic control⁶⁶, screen for metabolic syndrome in non-overweight/obese subjects⁶⁷ the current cardiometabolic risk factors and over 10 years among people with type 2 diabetes mellitus^{68,69}, and all-cause and cardiovascular disease (CVD)-related mortality in this group of patients⁷⁰. Interestingly, anthropometric indices may mitigate the burden of T2DM and its complications in low- and middle-income countries¹².

This study's limitations include reliance on experience from a single center and the fact that diet, exercise, stress, treatment, and genetics are known confounders of anthropometric measures of obesity and T2DM. Moreover, we used HbA1c alone as the diagnostic tool for T2DM. Additionally, methodological constraints, such as selection bias risk and potential confounding factors, may bias associations.

Conclusion

In the current study, BMI, WC, and WHtR were poorly predictive of T2DM in adult Sudanese. Combined indices or alternative metrics (e.g., body adiposity index or body roundness index) might provide better discrimination. Moreover, we encourage further studies to evaluate these anthropometric indices in Sudan to detect T2DM.

Data availability

The datasets generated and/or analyzed during the current study are not publicly available (because the manuscript is still under peer review), but are available from the corresponding author upon reasonable request.

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Author contributions

I.R.M. and I.A. conceived the study; I.R.M., O.E.O., and I.A. supervised the work, guided the analysis, and critically reviewed the manuscript; O.E.O., A.M.K., and I.A. prepared the analysis plan, performed the data analysis, and wrote the first draft of the manuscript. All authors reviewed and approved the final manuscript.

Declarations

Competing interests

The authors declare no competing interests.

Ethics approval

This study complies with the Declaration of Helsinki. The Ethics Committee of the Health Authority in Almatamah, Sudan, provided ethical approval of the study (reference number #9, 2021). Written informed consent was obtained from each participant.

Additional information

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