

The association between various biological, behavioural and psychosocial factors and type 2 diabetes mellitus in Africa: a systematic review and meta-analysis

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Abstract

Background

Type 2 diabetes mellitus (T2DM) is a significant public health concern in many African countries. While the determinants underpinning T2DM are likely to be Africa-specific, knowledge of these risk factors is largely derived from developed countries. This is the first systematic review and meta-analysis to include biological, behavioural and psychosocial risk factors for T2DM in Africa.

Methods

Relevant scientific databases were searched, and data were extracted from 66 studies. Fifty-nine studies reported unadjusted data and were analysed using Comprehensive Meta-Analysis (CMA) software version 2.0. The Odds ratios (OR) and their 95% CIs for the associations between BMI indices (overweight, obesity), central obesity (waist circumference, waist to hip ratio), behavioural (alcohol, fruit and vegetable consumption, smoking), physical inactivity, and psychosocial factors (stress, anxiety, and depression) and T2DM were calculated using a random-effect model. Moderator effects of age, language-spoken, sub-regions and urban/rural location was assessed.

Results

A number of risk factors were associated with T2DM including, BMI-based definitions of obesity [OR = 3.22, 95% CI: 1.90, 5.38], overweight [OR = 2.22, 95% CI: (1.90, 2.58)], or overweight/obesity [OR = 2.58, 95% CI: (1.76, 3.78)]; Central obesity as measured by waist circumference [OR = 2.51, 95% CI: (1.86, 3.37)], or waist to hip ratio [OR = 2.03, 95% CI: 1.51, 2.72]; psychosocial factors which includes stress [OR = 2.2, 95% CI: (1.46, 3.31)], depression [OR = 2.42, 95% CI: (1.14, 3.31)] and anxiety [OR = 2.05, 95% CI: (1.0, 4.18)] and physical inactivity [OR = 1.88, 95% CI: (1.53, 2.23)]. Current smoker [OR = 1.13, 95% CI: (0.84, 1.49)], alcohol consumption [OR = 1.10, 95% CI: (0.82, 1.47)] and inadequate fruit and vegetable consumption [OR = 0.81, 95% CI: (0.57, 1.16)] were not associated with T2DM. Locality (urban/rural), language spoken, and sub-region (East/West Africa) did not significantly moderate the associations between the risk factors and T2DM.

Conclusion

Obesity (defined by BMI) is most strongly associated with T2DM. Overweight, waist circumference and waist to hip ratio, physical inactivity, psychosocial risk factors defined as stress, depression, anxiety were all significantly associated with T2DM. These findings add novel meta-analyses of associations between diverse individual risk factors and T2DM within the African context.

Background

Introduction

Diabetes Mellitus (DM) is a major contributor to global mortality and morbidity. Since the 1980s, Africa has seen an emergence of T2DM as an important non-communicable disease (NCD) increasingly threatening the health, political and socio-cultural framework of the continent's population [1, 2]. Across Africa, T2DM prevalence varies significantly within and between countries, geographical location (e.g. urban vs rural) [5, 6] and sub-regions (e.g. East vs West Africa) [4], as well as countries of common or different spoken language (e.g. French vs. English speakers) [3]. Comparative figures show that while the prevalence of diabetes in other continents is expected to increase by between 15% (Western Pacific) and 84% (South East Asia) from 2017 to 2045, the prevalence in Africa is predicted to increase by 156% within the same period [4]. As such, Africa is on the path to bear the most significant burden of T2DM epidemiology including Disability-Adjusted Life Years (DALY), premature death and mortality in the decades ahead [5].

In sub-Saharan Africa (SSA), 69% of people with T2DM aged 20 to 79 are undiagnosed [6]. This figure is substantially greater than the global figure of 50% [6]. As such, the majority of Africans with T2DM are diagnosed only after presenting with substantial health complications. This delay in diagnosis, coupled with a lack of access to high-quality healthcare, results in loss of productivity, morbidity, premature death [7], and mortality [8]. The Global Burden of Disease study suggests that all-age total DALY lost due to NCDs including diabetes in SSA, increased by 67% between 1990 (90.6 million [95% UI 81.0–101.9]) and 2017 (151.3 million [133.4–171.8])

[5]. Broader knowledge of the relationship between modifiable risk factors and T2DM underpinning this burden and the escalating trend is based predominantly on studies involving Europid and other non-African populations [9, 10]. Previous studies suggest that the relationship between risk factors such as body weight indicators (e.g. obesity & waist circumference) and T2DM varies across different populations [9, 11–13]. The unique racial composition, culture, socioeconomic factors, dietary patterns, political structures, geography and environment in Africa mean that Africa-specific studies of risk factors for T2DM are important. Understanding the African-specific relationships between potentially modifiable risk factors (biological, cultural, behavioural and psychosocial) and T2DM is of critical importance to primary prevention efforts [2, 14, 15].

Although several primary Africa-specific epidemiological studies have been conducted investigating associations between modifiable risk factors and T2DM [16–35], to date no reviews, to our knowledge, have summarised these studies using a systematic and meta-analytic methodology. This is despite the existence of several Africa-specific T2DM systematic and narrative reviews [36–40], including limited meta-analyses [40–46]. These reviews, however, focused primarily on the epidemiology of T2DM burden (e.g. prevalence, incidence) and their outcomes (morbidity and mortality) [36–40, 47–52] and not on the magnitude of associations. This is the first systematic review and meta-analysis, to our knowledge, to focus on the relationships between a range of biological, behavioural and psychosocial risk factors and T2DM in Africa. This study also considers the role of moderators including age, sex, sub-region (east, west, northern or southern), language spoken, and geographical location (rural vs urban).

Methods

Design

This study follows the Preferred Reporting Items for Systematic Reviews and Meta-Analysis (PRISMA) guidelines in reporting of this systematic review and meta-analysis [53].

Figure 1 depicts the PRISMA flow chart. The protocol for this study has been previously published [3] and is registered with PROSPERO (Registration number CRD42016043027).

Search Strategy

We conducted a database search in English of published quantitative data only. The initial search was conducted in November 2017 with an updated search in February 2019. The search included studies that focused on diabetes and a range of biological, behavioural and psychosocial risk factors within the African context. Please see Additional file 1 for a list of search terms. The search covered scientific databases including Global Health, PsycINFO, CINAHL, MEDLINE, Psychology and Behavioural Science. Reference lists of included articles were searched for additional articles.

Inclusion and Exclusion Criteria

The study considered articles that reported empirical research, were conducted among populations living in Africa and published in English from 1 January 2000 to 16 February 2019. We included quantitative studies from cross-sectional, longitudinal, cohort and case-control that reported associations between T2DM as an outcome measure and the risk factors of interest. Studies were included if T2DM was measured following the 1999 World health organisation (WHO) and the American Diabetes Association (ADA) criteria. The risk factors examined include body weight indicators which include BMI indices [overweight (OV), obesity (OB) and overweight/obesity (OV/OB)], and central obesity indices (waist circumference (WC) and waist to hip ratio (WHR)). Others include physical activity (PA), current alcohol consumption (ALC), current smoking (SMO) and psychosocial risk factors (stress, anxiety, and depression).

We excluded studies with outcome measure as type 1 diabetes, impaired glucose tolerance, impaired fasting glucose and gestational diabetes.

For the independent risk factors, we excluded socioeconomic status as a potential risk factor due to its direct link to both non-modifiable (e.g. age, sex, and ethnicity) and modifiable risk factors (e.g. SMO, ALC, diet, and PA) as opposed to T2DM [54]. We also excluded hypertension and dyslipidaemia as T2DM may causally affect hypertension and dyslipidaemia whereas the opposite relationships are unlikely to be causal [55, 56].

Reviews, editorials, commentaries, and letters were excluded. Studies that reported relevant risk factors and T2DM, but not the associations between them, were excluded. Associations with risk factors treated as the outcome measures in multivariate analyses

adjusted for various covariates were also excluded.

Screening

The search yielded 39,561 studies and results were imported into Endnote Version 9. Duplicates were removed and 30,713 studies remained. The first reviewer (AI) assessed study eligibility for inclusion by initially screening the titles and abstracts. For an article to pass the first stage of screening, the title, and abstract needed to mention diabetes and any of the risk factors. Five hundred and four studies were retained, and full text obtained for the second stage. Another 8 articles were found in the reference lists. At this stage, the articles were further considered if three reviewers (AI, YP, and AC) considered that their abstract showed that the paper may contain quantitative data and an association between the risk factors and T2DM. The observed average agreement between reviewers on the screening at this stage was 95.9%. After completing this stage, a total of 177 studies were retained for the third stage and screened against all criteria including an appropriate measure of associations by the first reviewer and results confirmed by a second reviewer (YP). Final decisions to exclude studies at this stage were discussed by two reviewers (AI and YP) and disagreements were resolved by consensus with third (AC) and fourth (CS) reviewers. The agreement between reviewers at this stage was 97.5% with 66 studies remaining for data extraction [16–22, 24, 25, 27–34, 57–94] and including both unadjusted and adjusted data (e.g. adjusted for covariates such as age and family history). Only unadjusted data were included in the meta-analysis.

Data extraction and management

The final 66 studies were reviewed, extracted and coded by the one author (AI), and extracted data were double-checked by YP. We developed a Microsoft Excel spreadsheet and an accompanying manual for this purpose. We then extracted data from studies that reported an unadjusted bivariate association between the risk factors and T2DM as well as studies that adjusted for different covariates. Information extracted from each article includes study design, number of studies (*k*), year of publication, study participants (*n*), risk factors (e.g. OB) and associations with T2DM for effect size calculations [95]. Others include age, geographical locations (urban, rural), countries and sub-regions, language spoken, sample size, and odds ratio (OR) with confidence intervals (CI).

The definition of T2DM was based on WHO or ADA criteria, and included a fasting blood glucose (FBG) of ≥ 7.0 mmol/L (126 mg/dL), or 2-hour glucose level of ≥ 11.1 mmol/L (200 mg/dL) during a 75 g oral glucose tolerance test (OGTT) or the glycated haemoglobin (HbA1c) value of $\geq 6.5\%$ [96].

Definitions of smoking, alcohol and fruit and vegetable consumption adopted for this study have been used extensively in previous reviews [97–100] and include: Current smoking status (yes or no, including former smokers. Current alcohol intake (either i. yes or no; or ii. current drinkers vs. never/non-drinkers). Fruit and vegetable consumption (inadequate vs adequate consumption) and physical activity (vigorous vs low physical). For psychosocial risk factors, the reference group was no stress, depression or anxiety vs. currently having these conditions. Due to the small number of studies for these individual risk factors (stress, anxiety, depression) they were aggregated into a summary variable for psychosocial risk factors.

Data Integration, coding and analysis

The metrics used in this meta-analysis in measuring associations between the risk factors and T2DM were ORs and 95% CIs, employed as a measure of the effect size. Other metrics were converted to ORs including cross-tabulation (2×2) of events and non-event (T2DM/ risk factor \times T2DM/no risk factor) as well as Chi-squared. Comprehensive Meta-Analysis (CMA) was used for all conversion, calculations as well as coding the data [101]. Where ORs were lower than 1 (i.e. risk factor was protective), ORs were reverse coded by calculating 1/ORs for consistency and ease of interpretation.

Variation in sample sizes was accounted for by calculating weighted effect sizes, giving larger samples more weight. Most studies provided multiple homogenous associations between risk factors and T2DM which were not independent. In resolving the issue of dependence, we used the averaging method to calculate a single effect size per study [102] (i.e., the shifting-unit-of-analysis approach [102]). This approach not only ensured that each study contributes exactly a single independent association per analysis but was more advantageous in retaining as much data as possible [102, 103]. Since the study aims to generalise the findings, the random effect model was used to calculate the overall effect sizes. For moderator analysis, mixed-effect models were a more conservative approach to test for different moderator levels. Moderation analyses were conducted separately for all risk factors.

Heterogeneity and moderator analysis

Q and I^2 statistics were used to assess effect sizes of study heterogeneity [104]. The Q statistics measured the existence of heterogeneity, while the I^2 statistics assessed the percentage of study variability due to heterogeneity instead of by chance [104, 105]. I^2 values of approximately 50% and $\geq 75\%$ were considered moderate and high heterogeneity respectively [104, 105]. Moderation analysis was used to explain the heterogeneity and interaction effects of the risk factors and T2DM. The strength and directions of these associations were tested using moderators such as study participants' age, sub-regions (East/West Africa), Locality (urban/rural dwelling) and lingua franca (French or English speakers). A moderator variable was analysed if it had at least two levels that were examined in five or more studies [95]. This is based on the established minimum threshold widely used in various meta-analyses [95, 104, 106, 107].

Vote-counting analysis

This study utilised unadjusted data only and was complemented by summarising the adjusted data using the vote-counting methods. Vote counting is a method of synthesising evidence from multiple estimates by comparing the number of studies with positive results against the number of studies with negative results. However, it does not take into consideration sample size, study quality and effect size [108]. The most common variables adjusted for in this study included hypertension, IFG, age, family history of diabetes, gender, socioeconomic status, education, and marital status [109].

Publication bias analyses

Using the CMA program, we assessed publication bias in three different ways. First, we checked for bias by producing funnel plots and physically examined any evidence of symmetry. Second, we examine the intercept for significance of the sample of studies for statistical evidence of bias using Egger's test [110]. Thirdly, due to the propensity to publish studies with statistically significant results relative to the non-significant ones, thus "the file drawer problem" [111], the Rosenthal's Failsafe number was used to calculate the number of non-significant unpublished (or missing) studies needed to be included to the meta-analysis to change statically significant results to non-significant one [112]. The criterion suggests that, the value of the Failsafe N should be greater or equal to 5 times the number of included studies in a meta-analysis. The Rosenthal's failsafe N was also used to decide whether the effect is an artefact of bias or by chance. The Duval and Tweedie trim and fill methods were then used to adjust for detected bias or unpublished (or missing) papers to assess what the effect size would have been given no bias existed [113, 114].

Results

Descriptive

An initial search generated a list of 30,713 publications after duplicates were removed. A total of 66 studies met the inclusion criteria, with 59 of them reporting unadjusted data and hence included in the analyses. Tables 1, 2, and 3 provide descriptive statistics of study, risk factor, and participant level characteristics respectively. The overall sample size was 181,204 participants. There was a high variation in sample sizes across the 66 studies. Study sample sizes ranged from 90 to 45,767 participants, with 37% of articles reporting a sample between 50–500, while 20% of studies had a sample size between 501–1000 and 35% included a sample size between 1001 to 5000. Nine percent of studies had a sample size between 5001- 45,767. The included 66 studies were comprised of 328 individual unique associations. Of these associations, 206 were unadjusted from 59 studies. Two articles reported two different studies each in one paper [20, 93]. These were comparative studies from multiple countries in Africa. We differentiated these studies by adding study 1 and 2 to each reference. The 66 articles were published between 2000 and 2019. Of these studies, 35 articles were published between 2016 to 2019 (53%), 23 studies between 2011 to 2015 (35%), and 8 studies between 2000 to 2010 (12%). All studies were published in academic peer-reviewed journals. Many of the studies were conducted among West African populations (39%). Of these, 21% were published in Nigeria, and others from Ghana (12%), Senegal (5%), and Burkina Faso (2%). This is followed by East Africa, where 38% of the total studies were published. Among these countries, 11% were from Ethiopia, or otherwise Uganda (8%), Kenya (9%), Malawi (2%), Mozambique (2%), Tanzania (5%), North Sudan (2%) and South Sudan (2%). The majority of the studies used a cross-sectional design (86%). Only 4 studies (6%) [21, 66, 79, 115] reported longitudinal data and 6% were case-control studies [16, 61, 116, 117]. More studies were conducted in urban areas (41%) compared to rural areas (21%), with 33% of studies conducted in both geographical areas.

Table 1
Study level characteristics

Article Level	Groups	Number of article reporting	Percentage
Sample Size	50–500	24	36.6%
	501–1000	13	19.7%
	1001–5000	23	34.8%
	5001–45,767	6	9.1%
Year of publication			
	2000–2010	8	12.1%
	2011–2015	23	34.8%
	2016–2019	35	53.0%
Type of publication			
	Academic journal	66	100.0%
Sub-regions			
West Africa			
	Nigeria	14	21.2%
	Ghana	8	12.1%
	Senegal	3	4.6%
	Burkina Faso	1	1.5%
Central Africa			
	DRC	3	4.6%
	Cameroon	2	3.0%
East Africa			
	Uganda	5	7.6%
	Ethiopia	7	10.6%
	Kenya	6	9.1%
	Malawi	1	1.5%
	Tanzania	3	4.6%
	Mozambique	1	1.5%
	North Sudan	1	1.52%
	South Sudan	1	1.52%
North & Southern Africa			
	South Africa	5	7.6%
	Seychelles	1	1.5%
	Zambia	1	1.5%
	Morocco	1	1.5%
Multiple countries			
	Kenya, Ghana & Nigeria	1	1.5%

Article Level	Groups	Number of article reporting	Percentage
	Uganda and Tanzania	1	1.5%
Language spoken			
	English	46	69.7%
	French	11	16.7%
	Amharic	7	10.6%
	Arabic/English	2	3.0%
Study type/design			
	Cross-sectional	60	90.9%
	Longitudinal	6	9.1%
	Case Control	4	6.1%
	NR	1	1.5%
Sex			
	Male and female	63	95.5%
	Female only	1	1.5%
	Not reported	2	3.0%
Age			
	Adults (18 and above)	53	80.3%
	Mix (12 and above)	12	18.2%
Urban/Rural			
	Both	22	33.3%
	Urban	27	40.9%
	Rural	14	21.2%
	NA	3	4.6%

Table 2
T2DM and risk factor level characteristics

T2DM and Risk Factor article Level	Groups	Number of articles reporting	Percentage
T2DM Measure			
Biologically Screened		66	100.0%
	FPG/RBG	44	66.7%
	OGTT	2	3.0%
	HBA1c	5	7.6%
	Glucose oxidase	3	4.5%
	Mix [Screened (FPG)] and self-report]	12	18.2%
T2DM Definitions			
	WHO only	45	80.4%
	WHO/ADA	4	7.1%
	ADA only	7	12.5%
Risk Factors			
	Overweight (BMI)	48	16.2%
	Obese (BMI)	46	15.5%
	Overweight/Obese (BMI)	15	5.1%
	Obese (WC)	34	11.4%
	Obese (WHR)	27	9.1%
	Physical inactivity	27	9.1%
	Alcohol	32	10.8%
	Smoking	40	13.5%
	Fruit & Vegetables	7	2.4%
	Fruit only	7	2.4%
	Vegetables only	4	1.3%
	Stress	2	0.7%
	Anxiety	2	0.7%
	Depression	4	1.3%

Table 3
Participant level characteristics

Participant level	Number	Percentage
Male	72,957	41.2%
Female	104,062	58.8%
Mix (12 and above)	30,105	17.0%
Adults (18+)	146,734	83.0%
Not reported	5239	3.0%
English	149,928	84.8%
French	24125	13.6%
Other	5750	3.3%

Table 2 shows risk factor and outcome level characteristics. Reported in the 66 studies, almost all 181, 204 participants were biologically screened for T2DM. Forty-three (65%) of these studies used FBG/RBG [16, 17, 19, 20, 22–30, 34, 57, 58, 60, 61, 64–66, 68, 69, 75, 77–82, 84–89, 91, 116–122]. Twelve (18%) studies reported diabetes diagnosed by both FBG test and self-report [18, 20, 21, 32, 67, 73, 92, 93, 115]. Five studies (8%) used HBA1c [83, 90, 123–125] while 3 (6%) studies used glucose oxidase test [31, 63, 94], and 2 (3%) used OGTT test [71, 72].

Among all the risk factors body weight indicators (BMI-OV/OB, BMI-OB, WC, and WHR) were most commonly reported with BMI-OV (16%) being the most reported overall. This was followed by BMI-OB (15%) and BMI-OV/OB (5%), WC (11%), and WHR (9%). Lifestyle risk factors reported included smoking (13%), alcohol consumption (11%) and fruit and/or vegetable consumption (2%). Further, 27 studies (9%) reported physical activity with 18 of these (67%) using the Global Physical Activity Questionnaire (GPAQ), followed by the General Practitioner physical activity questionnaire (GPPAQ) (7%), and 15% used self-reported and occupation assessment. Three percent of the studies examined psychosocial factors (depression, stress, and anxiety) utilising various instruments (please see Additional file 2). Five (8%) studies [26, 81, 90, 125, 126] reported a single association, while 61 (92%) reported associations with multiple risk factors. Table 3. shows participants' characteristics across all studies. All studies reported the sex and age of participants. Of the 181,204 participants, 41% were males and 59% were females, 83% (18 years and above) of the participants were adult and 17% were young adults (12 years and above), 85% were English speakers and 14% were French speakers.

Meta-analysis

Effect sizes (OR) by risk factors

The meta-analysis includes 59 studies reported in 57 papers. It comprised of 206 unadjusted associations between modifiable risk factors and T2DM. Table 4 presents the mean of weighted effect size results. This includes three main groups: overall body weight indicators (i.e. BMI indices and central obesity indices), psychosocial factors, and lifestyle risk factors. Among these three groups, the overall body weight indicators had the largest mean weighted effect size (OR = 2.44, 95% CI: [2.10, 2.84] k = 45), followed by psychosocial factors (OR = 2.15, CI: [1.51, 3.06] k = 4). Lifestyle risk factors were not significantly associated with T2DM.

Table 4
Effect sizes for associations between combined and individual risk factors and T2DM

Risk factor group	Risk factors	OR	Lower CI	Upper CI	Z-value	P-value	k	Q-Value	P-value	I-squared
Body weight indices (BWI)	BMI-OB	3.22	2.73	3.80	13.89	< 0.001	29	62.85	< 0.001	55.45
	BMI-OV/OB	2.58	1.76	3.78	4.85	< 0.001	12	53.31	< 0.001	79.37
	BMI- OV	2.22	1.9	2.58	10.18	< 0.001	29	70.88	< 0.001	60.5
	Combined effect(BMI indices)	2.59	2.23	3.00	12.53	< 0.001	41	114.67	< 0.001	65.12
	Obese (WC)	2.51	1.87	3.37	6.09	< 0.001	21	152.8	< 0.001	86.91
	Obese (WHR)	2.03	1.51	2.72	4.73	< 0.001	16	43.84	< 0.001	65.78
	Combined effect (Central Obesity Indices)	2.24	1.80	2.80	7.16	< 0.001	27	115.51	< 0.001	77.5
	Overall effect BWI	2.44	2.1	2.84	11.47	< 0.001	45	188.48	< 0.001	76.66
Psychosocial risk factors (PsyRF)	DEP	2.42	1.14	5.16	2.29	0.020	4	5.22	0.160	42.48
	STRESS	2.2	1.46	3.31	3.8	< 0.001	2	0.57	0.450	0.00
	ANX	2.05	1.00	4.18	1.96	0.050	2	0.53	0.470	0.00
	Overall effect PsyRF	2.15	1.51	3.06	4.25	< 0.001	4	3.97	0.410	0.00
Lifestyle Risk factors (LsRF)	ALC	1.05	0.90	1.23	0.66	0.510	22	76.98	< 0.001	72.72
	SM	1.12	0.84	1.49	0.75	0.450	28	174.64	< 0.001	84.54
	FR&VE	0.81	0.57	1.16	-1.15	0.520	3	4	0.100	76.34
	Overall effect LsRF	1.10	0.91	1.27	0.84	0.400	31	131.5	< 0.001	77.19
	FR	0.70	0.47	1.05	-1.72	0.090	7	14.32	0.030	58.12
	VE	0.92	0.55	1.54	0.34	0.740	4	10.19	0.020	70.56
Physical inactivity (PA)	PA	1.88	1.53	2.32	5.94	< 0.001	15	32.46	< 0.001	56.87
OV-Overweight; OB-Obesity; OV-OB- Overweight or Obese; WC-Waist circumference; WHR-Waist to hip ratio; PA-Physical inactivity; AL-Alcohol; SM-smoking; FR-Fruit; VE-Vegetables; DEP-Depression; STR-Stress; ANX-Anxiety; BW- Overall Body weight; PsyRF- psychosocial risk factors; LsRF- Overall lifestyle risk factors.										

For the bodyweight indicators, the OR for combined BMI indices (BMI-OV, BMI-OV/OB & BMI-OB) OR = 2.59, 95% CI: [2.23, 3.00. k = 41] was also larger than for the combined central obesity indices (WC & WHR) (OR = 2.24, 95% CI: [1.80, 2.80] k = 27). Figures 2, 3 and 4 presents forest plots for the overall body weight indicators, combined BMI indices, and central obesity indices. The mean weighted effect sizes for individual body weight indicators were BMI-OB (OR = 3.22, 95% CI: (2.73, 3.80. k = 29), BMI-OV (OR = 2.22, CI: [1.9, 2.58] k = 29), BMI-OV/OB (OR = 2.58, CI: [1.76, 3.38] k = 12), WC (OR = 2.51, CI: [1.86, 3.37] k = 21) and WHR (OR = 2.03, 95% CI: ([1.51, 2.72], k = 17). Except for BMI-OB (see Fig. 5), forest plots are not presented for individual body weight indicators due to space limitations but are available from the authors upon request.

PA was significantly associated with T2DM (OR = 1.88, 95% CI: [1.53, 2.23], $k = 15$); Additional file 3. Similarly, psychosocial risk factors associations with T2DM ranged from OR = 2.05, 95% CI: (1.00, 4.18) for anxiety to OR = 2.42, 95% CI: (1.14, 5.16) for depression.

Publication bias analyses

The publication bias analysis shows that, apart from BMI-OV/OB, funnel plots for the BMI indices and the other risk factors were comparatively symmetrical. However, the intercept for Egger's regression test was significant for BMI-OV/OB ($p = 0.022$), suggesting potential bias. The non-significant intercepts for the other BMI indices include BMI-OB ($p = 0.640$) and BMI-OV ($p = 0.923$). That of central adiposity were WC ($p = 0.0657$), WHR ($p = 0.247$), lifestyle risk factors ALC ($p = 0.427$), SMO ($p = 0.201$), psychosocial risk factors ($p = 0.838$) and PA ($p = 0.332$). The Duval and Tweedie trim and fill methods were used to impute 2 studies for BMI-OV/OB resulting in changes to weighted mean effect size (see Additional file 4). This was adjusted from OR = 2.58, CI: (1.76, 3.78) to OR = 3.13 CI: (2.01, 4.87), which remained significant indicating a likelihood of bias, with no significant impact. Further bias analysis using the fail-safe N indicates that possible bias may exist among the lifestyle risk factors (ALC, SMO, F & V) as well as psychosocial factors.

Vote-counting analysis

Vote counting included 28 studies reporting adjusted data with 82 associations between the risk factors and T2DM. Individual body weight indicators made up of 23 (82%) studies with 48 (59%) associations. Of the 82 associations, 50 (61%) reported common covariates, including age, sex, family history, BMI and high blood pressure (HBP). Other covariates adjusted for included ethnicity, socio-demographic and urban-rural migration. 44 (92%) of the 48 associations for the body weight indicators were associated with T2DM with point estimates ranging from OR = 1.08, 95% CI: (0.50, 2.31) to OR = 10.35, 95% CI: (3.34, 32.04). Of these, 24 were statistically significant (55%). The other 20 (45%) associations were not statistically significant.

Among the BMI indices, BMI-OB reported a point estimate of 14 associations with T2DM from 12 studies with point estimate ranging from OR = 1.19, CI: (0.35, 4.05) to OR = 6.55, CI: 1.20, 35.8). Of this, seven were statistically significant with point estimates ranging from OR = 1.49, 95% CI: (1.26, 1.75) to OR = 6.55, CI: (1.20, 35.8). A total of 15 studies reported 17 associations for BMI-OV. Fifteen (83%) associations showed increased odds of T2DM with point estimates ranging from OR = 1.17, 95% CI: (0.61, 2.25) to OR: 3.50, 95% CI: (3.20, 3.80). Of this 6 were significant [OR = 1.35, 95% CI: (1.16, 1.57) to OR = 3.50, 95% CI: (3.20, 3.80)]. Among the central obesity indices, WC had 7 associations with T2DM. Five of them were statically significant with increasing odds of T2DM. The point estimate ranges from OR = 2.23, 95% CI: (1.13, 4.40) to OR = 10.35, 95% CI: (3.34, 32.04). In addition, WHR had 8 associations, all of which showed increased odds of T2DM. The point estimate ranges from OR = 1.08, 95% CI: (0.5, 2.31) to OR = 4.6, 95% CI: (1.9, 10.9). Four (50%) associations were significant with point estimates ranging from OR = 2.63, 95% CI: (1.76, 3.93) to OR = 4.6, 95% CI: (1.9, 10.9).

Further, 8 studies reported 11 associations between PA and T2DM. Of these associations, 8 (73%) showed increased odds of T2DM. The point estimates ranged from OR = 1.04, 95% CI: (0.54, 3.56) to OR = 4.78 to 95% CI: (1.16, 19.65). Four were statistically significant and ranged from OR = 1.60, 95% CI: (1.10, 2.30) to OR = 4.78 to 95% CI: (1.16, 19.65). Finally, for combined smoking, alcohol and fruit and vegetable consumption, a total of 13 studies reported 21 associations. Fourteen (63.6%) of them showed increased odds with T2DM with point estimates ranging from OR = 1.03, 95% CI: (0.68, 1.58) to OR = 66.9, 95% CI: (28.97, 154.49). Of these five (35.7%) were statistically significant with point estimates from OR = 1.6 95% CI: (1.10, 2.40) to OR = 66.9 95% CI: (29, 154). Three of the significant associations were between smoking and T2DM and one was between alcohol and fruit and vegetable consumption respectively.

Moderator analysis

The results are shown in Table 5. Analyses were run per risk factor, only when each of the two moderator in question have 5 or more studies [107]. Sub-regions were significant in a pairwise analysis comparing effects for East Africans and the effects of West Africa for central obesity indices combined. The effect of East Africa was larger (OR = 2.80, $z = -25.55$, $p < 0.001$, $k = 9$) than the effect of West Africa (OR = 1.80, $z = 5.65$, $p < 0.001$, $k = 11$), $Q(Q1) = 5.22$, $p = 0.022$). However, for individual central obesity indices, WC was not moderated by sub-regions $Q(1) = 2.59$, $p = 0.11$. There was not enough data for WHR to enable such analysis. No other moderators qualified for analysis were significant.

Table 5
Effect Sizes (OR) for moderators of diabetes risk factors and T2DM

Moderators	OV	OB	OV/OB	Combined BMI Indices	WC	WHR	Combined Central Obesity	Overall BWI	AL	SM
Age										
Adults (>= 18 years)	2.20(21)	3.12(22)	—	2.65(30)	2.23(14)	2.27(9)	2.22(18)	2.51(33)	—	1.12
Mix (12 and above)	2.29 (8)	4.16(7)	—	2.57(10)	3.66(7)	1.66(7)	2.31(9)	2.35(11)	—	1.5
Between groups Q	0.05	1.45	—	0.03	2.5	1.36	2.03	0.15	—	0.45
Urban and Rural										
Rural	2.06(6)	3.42(7)	—	2.87(9)	4.18(6)	1.72(6)	2.73(8)	2.40(10)	0.89(5)	—
Urban	2.52(12)	3.46(12)	—	2.51(15)	2.02(8)	3.20(6)	2.45(10)	2.06(17)	1.04(11)	—
Between groups Q	0.67	0.002	—	0.34	3.59	2.34	0.11	0.12	0.08	—
Regions in Africa										
East	2.34(11)	3.51(11)	2.14(5)	2.55(17)	3.22(6)	—	2.85(9)	2.44(17)	0.93(7)	1.20(9)
West	2.21(10)	3.23(10)	2.43 (5)	2.50(14)	2.13(8)	—	1.80 (11)	1.86(17)	1.10(12)	1.38(11)
Between groups Q	0.06	0.14	0.11	0.01	2.59	—	5.22*	1.72	0.64	0.12
Central	—	—	—	4.47(5)	—	—	—	3.30(5)	—	—
East	—	—	—	2.62(17)	—	—	—	2.44(18)	—	—
Between groups Q	—	—	—	1.66	—	—	—	1.26	—	—
Central	—	—	—	4.47(5)	—	—	—	3.30(5)	—	—
West	—	—	—	2.14(17)	—	—	—	1.86(21)	—	—
Between groups Q	—	—	—	3.17	—	—	—	3.17	—	—
Language spoken										
English speaking	—	—	—	2.44(28)	—	—	2.25(17)	2.32(29)	—	—
French speaking	—	—	—	4.20(5)	—	—	2.45 (6)	3.08(8)	—	—
Between groups Q	—	—	—	2.06	—	—	0.04	1.09	—	—
OV-Overweight; OB-Obesity; OV/OB- Overweight or Obese; BMI indices- (OV + OB + OV/OB); WC-Waist circumference; WHR-Waist to hip ratio; Combined Central Obesity-(WC + WHR); AL -Alcohol; BWI- Overall body weight indices;(OV + OB + OV/OB + WC + WHR); SM-smoking; Numbers inside parentheses are studies 'numbers k. — Not available due to insufficient number of studies (k < 5) in at least one level of the moderator. *p < 0.										

Discussion

We conducted the first systematic review and meta-analysis focusing on a range of biological, behavioural as well as psychosocial factors and their associations with T2DM in Africa. Consistent with other systematic reviews and meta-analyses [95, 127–129], publication outputs have increased over time [130]. The greatest number of studies were from West and East Africa. The present study

showed that all body weight indicators (both BMI and central obesity defined), physical activity and psychosocial factors were significantly associated with T2DM with associations of varying strength. Obesity (defined by BMI) was found to have the strongest association with increased odds for T2DM of more than 3-fold that of any other body weight measures including waist circumference.

These general observations are consistent with some [131], but not all [132] prior meta-analyses. Findings from the present study are consistent with the study by Vazquez et al. [131] among populations from Mauritius, USA (including African Americans), Asia, and Europe. They are also mostly consistent with the study by Kodama et al. [132] among European subjects, however that study did suggest a stronger association between waist circumference and diabetes than that between obesity (defined by BMI) and T2DM. Waist circumference is a stronger indicator of intra-abdominal visceral fat than BMI, and closely linked to insulin resistance and hyperinsulinemia [133]. Interestingly, our findings of a stronger association between BMI and T2DM has also been documented in other pathophysiological studies in African using methods such as dual energy X-ray absorptiometry [134, 135]. Studies in South African women suggest that, for the same BMI, African women have less central fat, but greater peripheral fat accumulation than Caucasian women [133–135]. Sumner et al. [136] also show that increasing waist circumference results in less visceral adipose tissue among African-American and African women than Caucasian women. Although these findings were predominantly from studies among women, they may partly explain why the strength of the association between obesity (defined by BMI) and T2DM is stronger in this study than waist circumference.

The independent role of central obesity in insulin resistance in populations of African descent must not be discounted, however. For example, while detailed examination of the overall body weight indicators shows comparatively larger effect size for the relationship between combined BMI indices (BMI-OV, BMI-OV/OB, and BMI-OB) (OR = 2.59) and T2DM compared to the combined central obesity indices (WC and WHR) (OR = 2.24), a strong and independent association between central adiposity and T2DM was evident [137]. This may suggest that either WC or BMI alone could be used as risk factors for T2DM among Africans. The use of a measuring tape alone to assess WC may be appealing in a setting where resources are minimal [138]. A further issue which may have affected our findings in relation to associations between waist circumference and T2DM is the high heterogeneity ($I^2 = 87$) of studies, which could not be explained by the moderation analysis. In addition, although meta-analyses have suggested a stronger association between BMI and diabetes in women than men [131, 139], this could not be assessed in the present study due to too few studies reporting findings separately for females and males.

Our analysis of physical activity indicates an almost two-fold increase in odds of T2DM for those who are inactive. After adjusting for publication bias, the magnitude remained the same. These findings are congruent with previous meta-analyses among populations from China, the USA, and Australia that have explored associations of medium to vigorous physical activity and T2DM [140]. While the present study focused on vigorous physical activity and T2DM, other reviews included alternative physical activity measures such as walking, leisure-time activity, resistance activity, occupational activities, low, moderate and vigorous-intensity activity [141]. These measures are important within the African context since physical activity patterns in Africa are somewhat different from industrialised countries [142]. Low to moderate activities are most prevalent within Africa [142]. Although beyond the scope of this paper, an in-depth comparative examination of these differences is required within the African context.

Psychosocial factors in this study were found to increase the odds of T2DM by more than 2-fold with a combined OR = 2.15. These findings are consistent with various meta-analytic reviews in non-African populations that explored relationships between psychosocial factors and diabetes [143]. A study by Smith et al. [97] in North Americans (includes Whites, African Americans, Hispanic and Chinese), European, Middle-Eastern and Asian populations found that anxiety increased the odds of diabetes by almost one and a half fold, while that of Ali et al. [144] among a population from the USA, Europe (the Netherlands, Finland, and Italy) and Iraq found that depression increased the odds of diabetes by almost one and a half fold. In the present study, however, the small sample size of depression ($k = 4$), stress ($k = 2$) and anxiety ($k = 2$) may have limited power to detect the association with T2DM. As such these findings should be treated as preliminary and interpreted with caution. Again, a fine-grained longitudinal study examining the individual psychosocial risk factors (stress, depression, and anxiety) in the African context is required given evidence of differential associations with T2DM.

Findings of the lifestyle risk factors showed that in this African sample, associations between fruit and vegetable intake, alcohol consumption, and smoking and T2DM were not significant. However, various studies among non-African populations have shown otherwise. A meta-analysis of longitudinal studies among populations from the United States, Japan, United Kingdom, Germany, Israel, and the Scandinavian countries, showed that active smoking is associated with an increased risk of type 2 diabetes [100]. Similarly, a meta-analysis of longitudinal studies among populations from Europe, USA, Australia, Korea and Japan showed a U-shaped relationship between alcohol consumption and diabetes for both men and women, with a greater protective effect of moderate

consumption observed for women [145]. The non-significant findings of the present study may be partly due to the simpler definitions used in many African studies. Assessing consumption levels based on a report of “Yes” or “No” is likely to not be precise enough to detect an effect, particularly where the association is not strong, or there is a U-shaped association [145, 146].

Studies on diet quality show that adherence to the appropriate diet can improve insulin sensitivity and glycaemic control. The current study had data on fruit only and vegetable consumption only, which are protective factors for diabetes. However, only two studies presented associations for fruit and vegetable intake combined, and no study had data on consumption of discretionary, typically ultra-processed, unhealthy foods, which are strongly related to higher body weight and diabetes [147, 148]. As such, we were unable to confirm the link between diet and T2DM among the African populations. There is a clear need for further studies (preferably longitudinal) assessing the complex association between diet and the incidence of Type 2 diabetes in Africa [149].

These findings were supported by the adjusted data synthesised by the vote-counting analyses, specifically with regards to the more significant results including all body weight indicators. Most moderation analyses were non-significant except for the combined central obesity indices which were moderated by locality (East and West Africa). However, no other moderator was significant. This implies that the risk of acquiring T2DM is independent of geographical location (urban/rural) and spoken language in Africa.

Strengths and Limitations

There are several limitations associated with this study that should be considered when interpreting the findings. First, only studies published in English were included and thus may under-represent studies predominantly from the Arabic, Portuguese and French-speaking countries in Africa. The use of only unadjusted data may constitute a limitation due to challenges posed by adjusted covariate for systematic reviews and meta-analysis [127]. However, previous systematic reviews and meta-analyses reported non-significant findings between unadjusted bivariate and those adjusted for different covariate [132, 150]. Further studies can also explore moderations using meta-regression. In this study, various diabetes definitions were used among the included studies. However, it is unlikely that these differences in definition would have had any impact on the results as the aim of this study was to determine the associations between the risk factors and T2DM, and not the optimal cut-point for each risk factor [138].

The absence of study quality assessments in this study is another potential limitation. Using a critical appraisal tool for meta-analytic studies exploring risk factors and T2DM have been rare, due to challenges posed by observational studies with diverse methodologies. It is also possible different study qualities from individual studies may affect the association between risk factors and T2DM. For example, the observed association between physical activity and diabetes in Africa may have been influenced by using different measurement approaches as well as other confounders used in individual studies. In particular, the different instruments used may constitute a limitation as in the case of psychosocial factors where no two studies used the same tool. Nevertheless, the influence of unknown residual confounders can also not be ignored in individual studies (particularly the lifestyle risk factors) as this may affect the results. There is an urgent need for quality data, particularly on lifestyle risk factors. In sum, there is urgent need of longitudinal studies with a probability sampling technique to systematically measure both risk factors and T2DM, particularly alcohol, smoking, psychosocial factors, unhealthy food, and a range of relevant risk factors in Africa.

Our study has many strengths which include the systematic nature of the review and the use of comprehensive meta-analytic methods, which have not been used to determine the strength of association between risk factors and T2DM in Africa previously. Finally, the large sample size of the study provides high power and precision in our estimates.

Conclusion

This study constitutes the first comprehensive systematic review and meta-analysis of the association between a range of biological, behavioural, and psychosocial risk factors and diabetes in Africa to date. These findings add novel meta-analyses of associations between diverse individual risk factors and T2DM. Risk factors including obesity (defined by BMI), overweight (defined by BMI), and overweight/obesity (defined by BMI); central obesity (defined by waist circumference and waist to hip ratio), physical activity, psychosocial risk factors (stress, depression, anxiety) were all significantly associated with T2DM. The present study shows that, unlike for studies of European populations, obesity, as defined by BMI, was the factor most strongly associated with T2DM.

This research provides an update on the associations between a range of factors and diabetes in Africa. This is a growing field of research as confirmed by the increased rate of publications over time. While this metanalysis has identified gaps in the literature, we

hope that this metanalysis provides renewed attention and novel directions for future studies in understanding modifiable risk factors and their role in tackling the escalating burden of T2DM in Africa [151].

Abbreviations

AL: Alcohol; ANX: Anxiety; BMI: Body mass index; CI: Confidence interval; CMA: Comprehensive Meta-Analysis; DALY: Disability-Adjusted Life years; DEP: Depression; FR: Fruit; FBG: Fasting blood glucose; GPPAQ: General Practitioner physical activity questionnaire; (GPAQ) Global Physical Activity Questionnaire; HBA1c: Glycated haemoglobin; IDF: International Diabetes Federation; IFG: Impaired Fasting Glucose; NCD: Noncommunicable disease; OB: Obesity; OGTT: Oral glucose tolerant test; OR: Odds ratio OV: Overweight; OV/OB: Overweight or Obese; PA: Physical activity; PRISMA: Preferred Reporting Items for Systematic Reviews and Meta-Analysis; PROSPERO: International Prospective Register of Systematic Reviews; PsyRF: psychosocial risk factors; RBG: Random blood glucose; SD: Standard deviation; STR: Stress; SM: smoking; SSA: Sub-Saharan Africa; T2DM: Type 2 diabetes mellitus; WHO: World Health Organization; VE: Vegetables; WC: Waist circumference; WHR: Waist to hip ratio.

Declarations

Ethics approval and consent to participate

Not applicable.

Consent for publication

Not applicable.

Availability of data and materials

All data used generated or analysed during this study are available from the corresponding author on reasonable request.

Competing interest

The authors declare that they have no competing interests.

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Author Contributions

AI and YP conceived and designed the study. AI curated, analysed and visualised the data and wrote the original draft of the manuscript. AI, YP, AC and CS were involved in screening of studies and reviewed the manuscript. AI edited the manuscript. All Authors read and approved the final manuscript.

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Figures

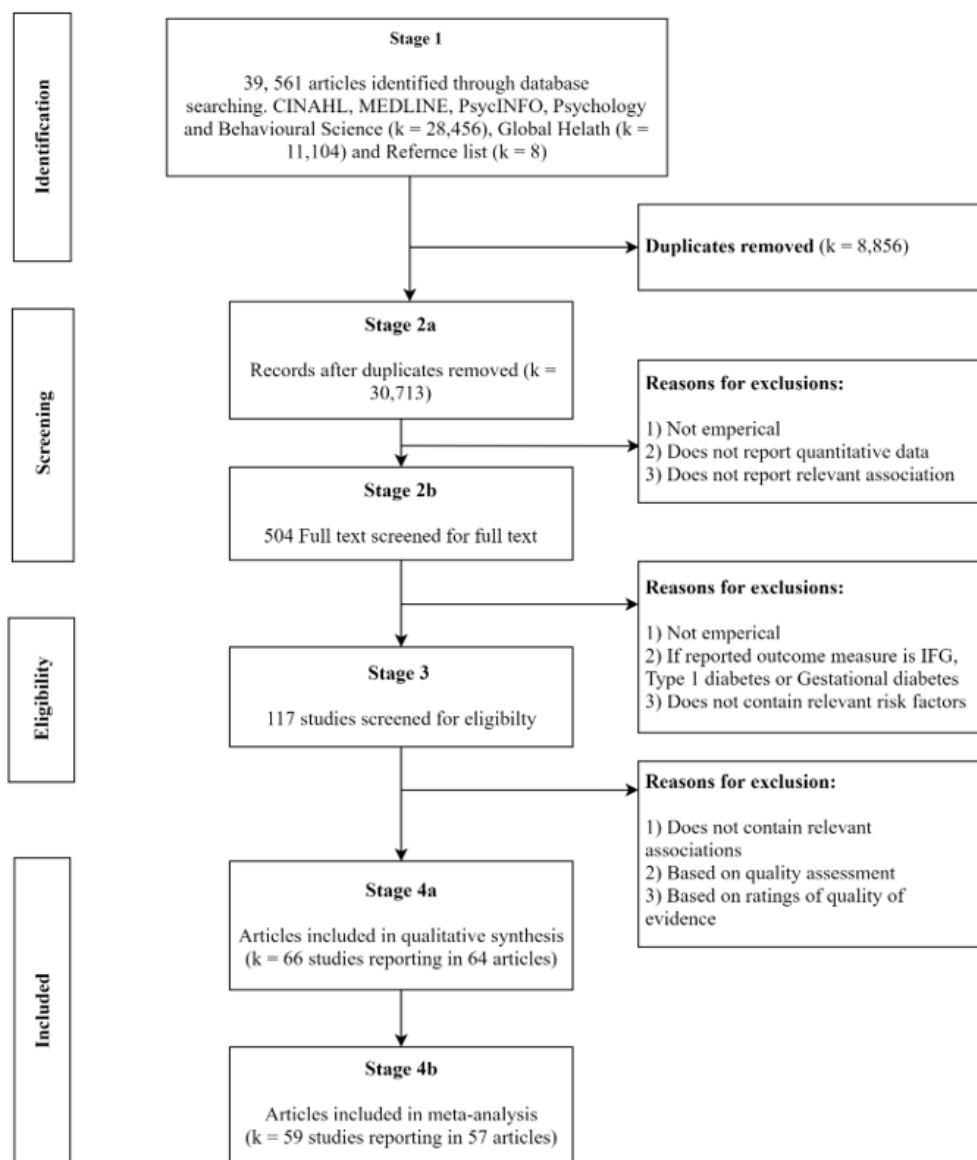


Figure 1

PRISMA screening process flowchart.

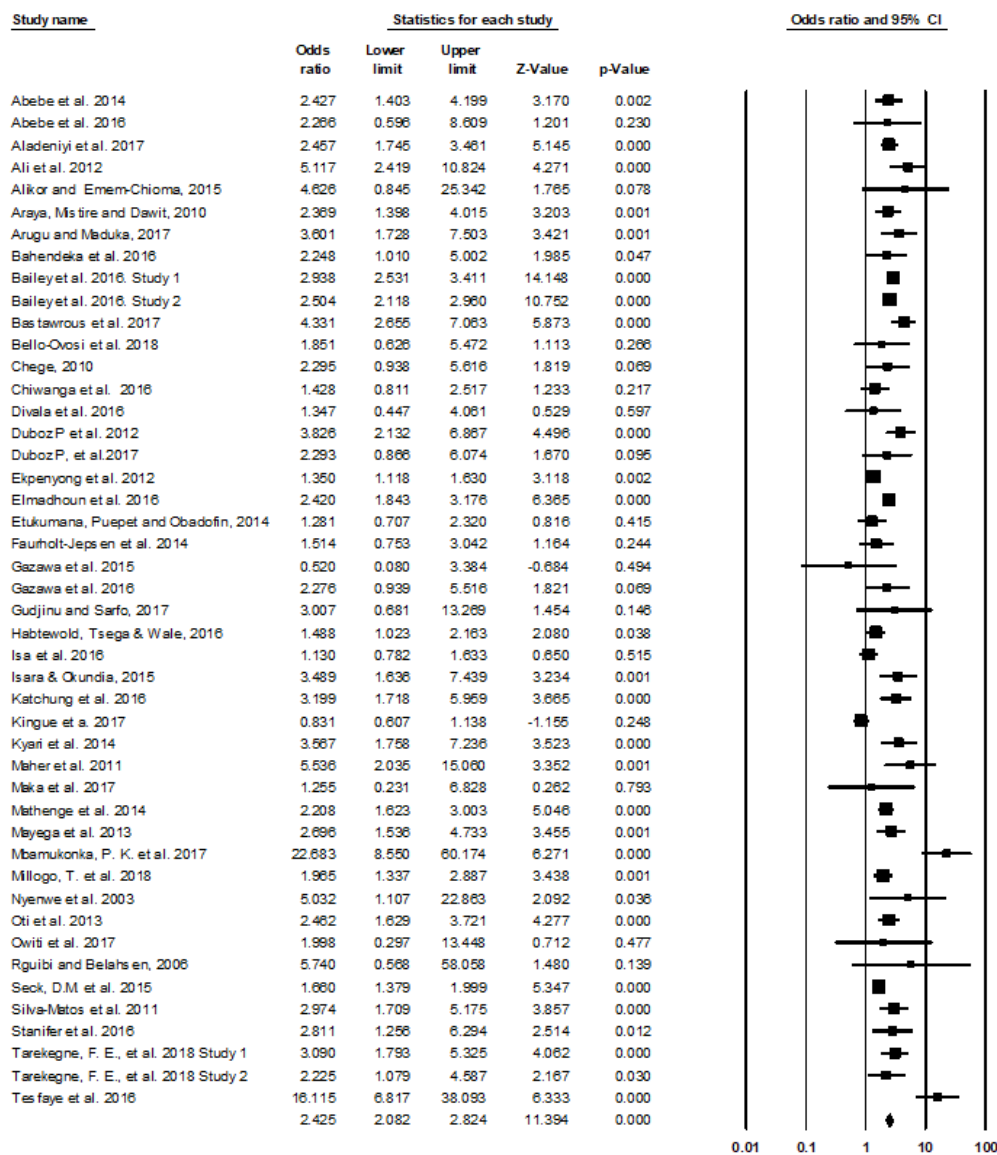


Figure 2

Forest plot of the effect sizes for overall body weight indicators included in the meta-analysis

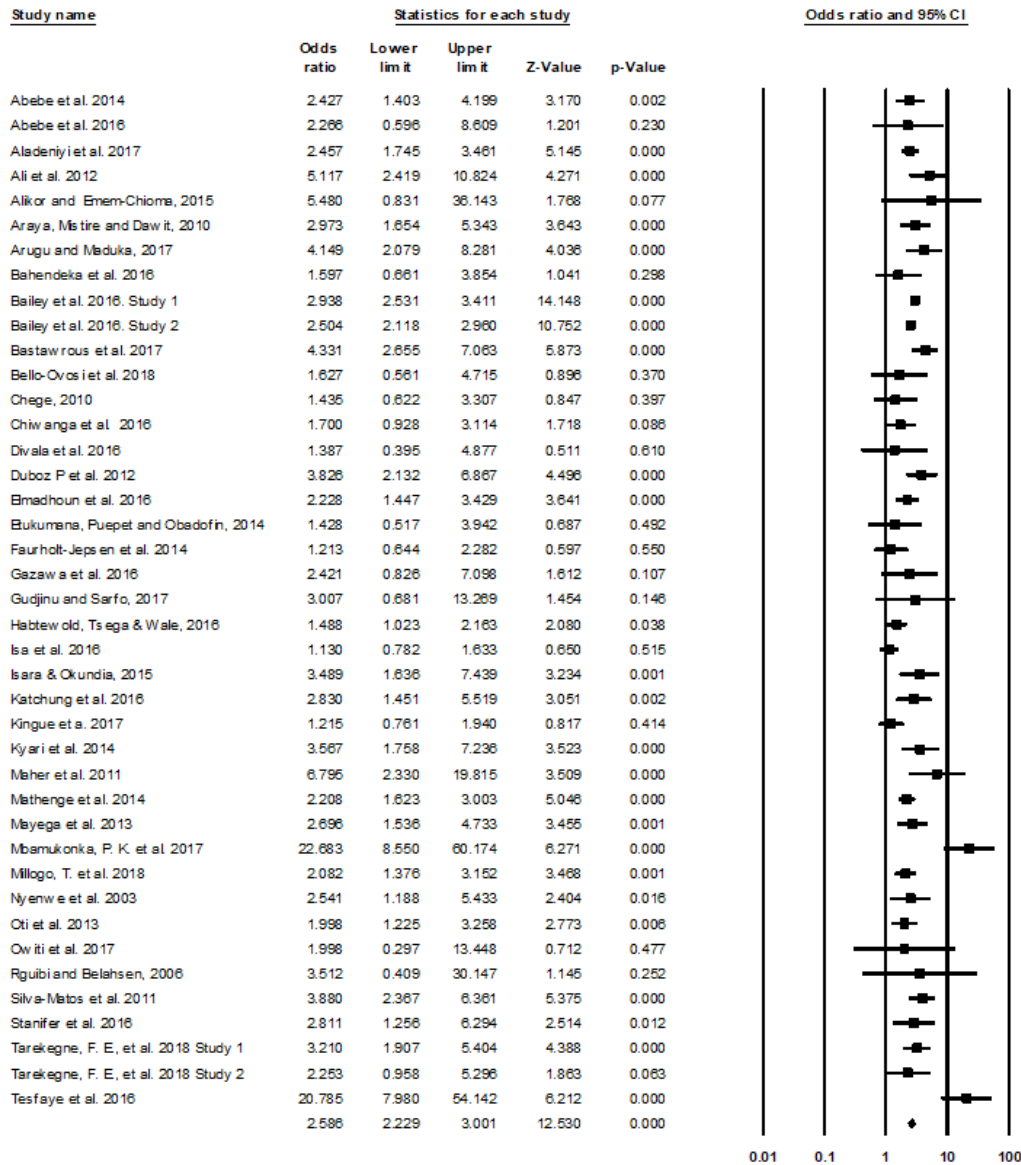


Figure 3

Forest plot of the effect sizes for combined BMI indices (BMI-OV, BMI-OV/OB, BMI-OB)

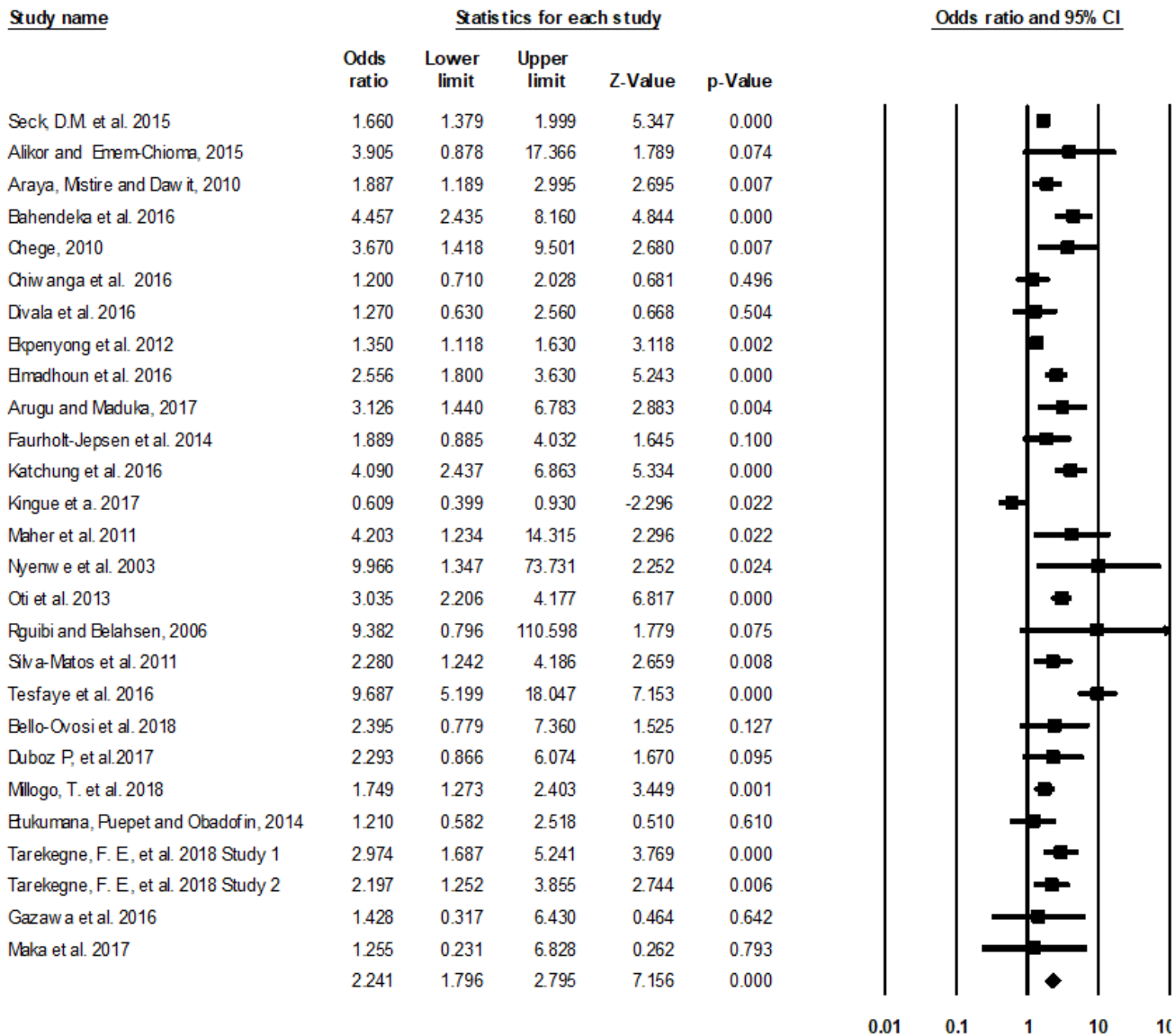


Figure 4

Forest plot of the effect sizes for Central obesity indices (WC & WHR)

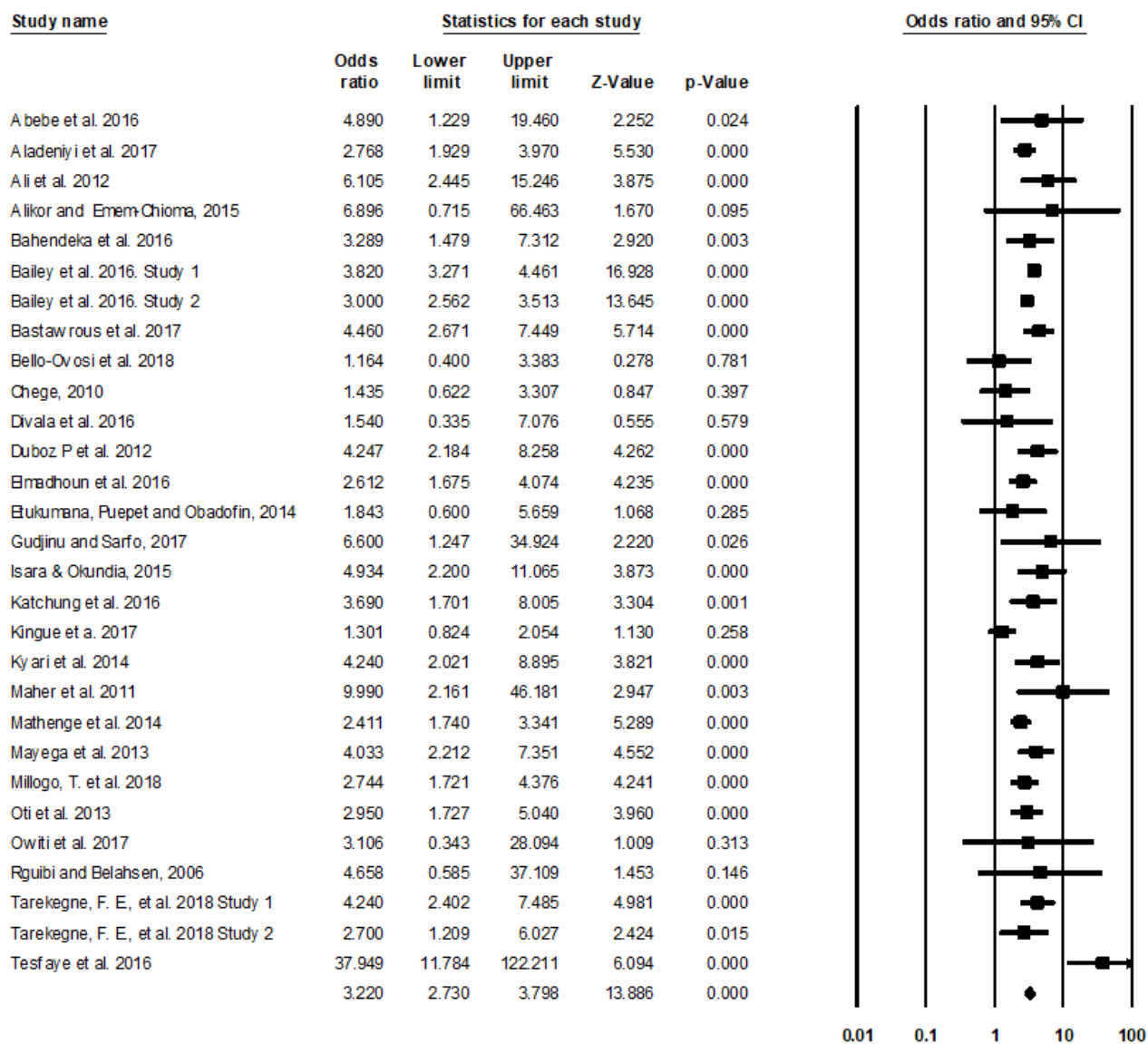


Figure 5

Forest plot of the effect sizes for individual studies: Obesity

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