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Article

Exploring New Tools to Risk Classification among Adults with Several Degrees of Obesity

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Abstract: The pandemic of obesity worldwide has been recognized as a very important challenge. Within its complexity the identification of higher risk patients becomes essential since it seems unsustainable trying to offer access to treatment to all people with obesity. Several new approaches have recently been presented as important tools for risk stratification. This research applied some of these tools in a cross-sectional study involving adults with obesity classes I, II, III and super obesity. The participants had their cardiometabolic risk profile assessed. The study included adults with obesity, aged 18 to 50 years (n=404) who were evaluated for anthropometric, body composition, hemodynamic, physical fitness and biochemical assessments. These variables were used to identify the prevalence of risk factors for cardiometabolic diseases according to the classes of obesity, by gender and age group. The results showed a high prevalence of risk factors, especially among the upper classes of obesity (BMI > 35 kg/m²) using single parameters as the waist circumference with almost 90% above the cut-off point. But there were also smaller numbers as the Glycated Hemoglobin whose prevalence was around 30%. Indexes like the atherogenic index of plasma (AIP) had the highest prevalence, with 100% of the male participants identified with increased risk for cardiovascular diseases.

Keywords: obesity; risk assessment; metabolic syndrome; AIP; HOMA-IR; risk stratification

1 Introduction

The highly complex etiology of obesity and its dynamic, encompassing genetic, physiologic, environmental, psychological, social, economic, and even political factors interacting in several ways promote and aggravate the obesity pandemic [1–3]. For this reason, it is hard to treat obesity effectively [4]. It is well known that the excess of adipose tissue, particularly the ectopic fat depots, are implicated in more than 200 complications of obesity and negatively impact the health of individuals affected [5].

The high and rising prevalence of obesity along with the associated health impacts represent a real challenge for the health and policy authorities, since it can be economically prohibitive to offer treatment access to all people in need [6–8]. Additionally, obesity is not necessarily synonymous with health risks, as the metabolically healthy obese (MHO) phenotype and also the fat-but-fit paradigm have also been documented [9,10].

In fact, there is consistent evidence about the prevalence of MHO with some studies presenting very high rates of that phenotype, like one with Brazilian women in which the prevalence of MHO

was around 70% considering the HOMA-IR and the NCEP-ATPIII criteria for metabolic syndrome [6]. Although, the prevalence of this phenotype can vary between $\approx 15\%$ and $\approx 30\%$ depending on the definition of Metabolic Syndrome (MetS), using the criteria of meeting 0 or 0–1 MetS components, respectively [10]. Studies have also revealed that the MHO is an unstable state, since an important part of the MHO subjects evolved to the unhealthy phenotype within some years [9,11]. The necessity to find better ways to diagnose people with obesity with the use of risk stratification categories is important to determine who is in high risk and in need of intervention strategies. The traditional criteria diagnostic of MetS requires the presence of three or more from the 5 components, including greater waist circumference (WC), dyslipidemia with high triglycerides (TG), low high-density lipoprotein cholesterol (HDL-C) levels, elevated blood pressure (BP) and impaired fasting plasma glucose (FPG) [12]. Because of its dichotomous or binary nature, some authors have proposed one option in which a continuous metabolic syndrome risk score can be used. This approach has the advantage of preserving the statistical power which is decreased when dichotomizing continuous variables. It also allows to measure in a more precise way the MetS risk along a continuum [12–14].

The one-size-fits-all approach should be avoided when programs to treat obesity are offered. The identification of subgroups whose risk profile is distinct is important to improve clinical practice [7,10]. It is also essential to make feasible offering access to treatment programs to those whose health risks are higher [3]. Hence, the main objective of this study was to assess the prevalence of traditional and recently developed risk factors assessing tools like the triglyceride glucose (TyG) index and related indexes, the continuous metabolic severity scores (MetSs) and the atherogenic index of plasma (AIP), related to the different obesity categories in a sample of Brazilian adults.

2. Materials and Methods

This descriptive study of cross-sectional design was carried out with 404 adults of both sexes, aged between 18 and 50 years, with obesity (Body Mass Index - BMI ≥ 30 kg/m²). Participants were selected to take part in the research project during the years of 2018, 2019, 2020 (first semester) and 2022 (second semester), in the Multidisciplinary Treatment of Obesity Program (MTOP), coordinated by the Multidisciplinary Obesity Studies Nucleus (NEMO) of the State University of Maringá (UEM) and Regional University Hospital of Maringá (HUM). Detailed description of the study can be found elsewhere [15]. In brief, eligible participants were invited to take part voluntarily in the study through dissemination in the local media (TV, radio, newspaper) and social networks (website and institutional email, Facebook). The interested participants took part in a pre-inclusion phase (Cardiometabolic Risk Assessment; CAR, divided into two steps) to confirm their eligibility to be involved in the study. In step 1, the conditions of eligibility were verified (age over 18 and under 50 and BMI over 30 kg/m²) and 774 people answered an anamnesis which included socioeconomic and health data. They had also evaluated: body mass, height, BMI, waist circumference (WC), and body composition by bioimpedance. Along with that their blood pressure (BP) and basal heart rate (HR) was measured. Finally, the physical fitness tests including the sit and reach for flexibility, the 30 seconds sit and stand for lower limbs resistance, the plank strength test for abdominal static resistance, and the six minutes' walk test (6MW) for cardiorespiratory fitness were applied.

After that process, a total of 404 people met the inclusion criteria, and were therefore considered eligible to participate in step 2, which included carrying out laboratory tests to verify the profile of cardiometabolic risk through the fasting measures of blood glucose, insulinemia, glycated hemoglobin, total cholesterol, HDL-c, LDL-c, VLDL-c, triglycerides and ultrasensitive C-reactive protein.

In order to determine the dosages of all those biochemistry variables, standard procedures were applied by specialized professionals from a private laboratory with quality control and an ISO certification.

Beyond the single parameters mentioned above, other surrogate measures of Insulin Resistance (IR) were determined by the Homeostasis Model Assessment (HOMA-IR) calculated as follows: $\text{HOMA-IR} = (\text{insulin} \times \text{glucose}) / 22.5$ [16,17]. The evaluation of homeostasis to verify the beta cells of

the pancreas was determined by the calculation (Homa-Beta): $20 \times \text{Insulin (iu/ml)} \div (\text{Glycemia} - 3.5)$, and the cut-off reference values were between 167 to 175 [18].

Triglyceride glucose (TyG) index calculated as $\ln [\text{fasting triglycerides (mg/dl)} \times \text{fasting plasma glucose (mg/dl)} / 2]$ [19]. We also calculated the product of triglyceride (TG) and fasting plasma glucose (FPG), the TyG index and also the TyG related to the adiposity status obtained by the equation (TyG/body mass index) and the TyG related to visceral adiposity by the ratio TyG by waist circumference [20]. It was also used for the risk assessment the atherogenic index of plasma (AIP), which is defined as the logarithm of plasma triglycerides to HDL-c ratio [21].

All procedures followed the requirements of Resolution 466/2012 of the National Health Council for research involving human beings which is made based on the principles of the Helsinki Declaration. Participants read and signed the Term of Free and Informed Consent agreeing to voluntarily participate in the research. The research was previously approved by the Permanent Committee of Ethics in Research of the State University of Maringá (Record nº 2,655,268).

The researchers involved in the assessments were all trained and followed standard procedures measuring anthropometric variables with the proper tools like height with a wall stadiometer (Sanny®), waist circumference (WC) with a flexible anthropometric tape (Medical Starrett-SN-4010 model, Sanny®) and body weight with a bioimpedance electric device (model InBody 520, Biospace®). Blood pressure was measured using an automatic arm monitor (model HEM-7113, Omron®). Blood collection and analysis for measuring blood glucose, high-density lipoprotein (HDL-c) and triglycerides was performed by qualified professionals in a private clinical analysis laboratory with quality certification between 7:00-9:00am with patients observing fasting for at least 8 hours.

To calculate the BMI, we used the formula: $\text{weight (kg)} / [\text{height (m)} \times \text{height (m)}]$ and the classification was based on the cutoff points of the World Health Organization (WHO, 2011). The WC measurement rating was also based on the WHO cutoff points, namely: WC >94 cm for men and >80 cm for women indicating increased risk of metabolic complications; whereas WC >102 cm for men and >88 cm for women indicated substantially increased risk of metabolic complications [22].

Blood pressure was classified according to the 7th Brazilian Guideline on Hypertension, as follows: normotension; systolic blood pressure (SBP) and diastolic blood pressure (DBP) $\leq 120/80$ mmHg; prehypertension: SBP between 121 and 139 and/or DBP between 81 and 89 mmHg and hypertension SBP ≥ 140 mmHg and/or DBP ≥ 90 mmHg [23].

Classification of fasting blood glucose followed the criteria of the Guidelines of the Brazilian Society of Diabetes 2017-2018, as follows: normoglycemia: fasting blood glucose < 100 mg/dL; pre-diabetes (or increased risk for diabetes mellitus): ≥ 100 to < 126 mg/dL; and established diabetes: ≥ 126 mg/dL (Oliveira, Montenegro Junior and Vencio 2017). The lipid profile was classified according to the 2017 Brazilian Guideline for Dyslipidemia and Atherosclerosis Prevention, as follows: high level of fasting triglycerides: ≥ 150 (mg/dL); low fasting HDL-c level < 40 mg/dL for men and < 50 mg/dL for women [24].

For data analysis, normality was verified using the Kolgomorov-Smirnov test. The average (\bar{x}) and standard deviation (SD) were used as descriptive statistics. To compare variables according to gender, unpaired t-tests were used. To compare the variables according to age group and level of obesity, a one-way ANOVA was used, with Bonferroni correction for multiple comparisons, and post-hoc tests were used to indicate between which groups there were differences. To correlate the variables, Pearson correlation coefficient was used. Analyzes were performed using the Statistical Package for Social Sciences (SPSS)® version 20.0 [25]. A significance level of $p < 0.05$ was adopted for all analyses.

3. Results

The participants of this study were 404 adults with obesity (Table 1), 85 men (21%) and 319 women (79%), aged between 18 and 50 years (mean \pm SD: 36.6 \pm 8.8 years). The BMI of the sample ranged from 31.3 to 77.2 kg/m² (mean \pm SD: 42.5 \pm 6.7). According to cutoff points of WHO, 76 (18.8%) had grade 1 obesity, 141 (34.9%) had grade 2 obesity, 160 (39.6%) had grade 3 obesity and 27 (6.7%)

had obesity grade IV or super obesity. Considering the degree of obesity, significant differences ($p<0.05$) were observed in all anthropometric/body composition variables, except height and in the relation lean mass/fat mass. At the same time, the classes/degree of obesity also presented significant differences related to the hemodynamic and physical fitness variables with DBP being higher in the group with obesity class IV and heart rate (HR) being higher in the group with obesity class III compared to class I ($p<0.001$). There were also significant differences in the total distance at the 6 minutes walk test with the group with obesity class IV presenting the lowest distance ($p<0.001$). Flexibility was also lower in the groups obesity class III and IV compared to class I and II ($p<0.001$).

Table 1. Cardiometabolic risk variables according to obesity classes among a sample of Brazilian adults (n=404).

Variables	Obesity class I (n=76)	Obesity class II (n=141)	Obesity class III (n=160)	Super Obesity (n=27)	p-value
Age (years)	41.66±8.75	39.50±9.10	38.27±8.77	37.15±8.32	0.028*
Height (m)	162.92±8.03	163.97±8.69	165.04±8.11	165.85±10.64	0.23
Body Mass (kg)	88.38±9.86 ^{a,b,c}	100.20±11.27 ^{a,d,e}	119.14±15.53 ^{b,d,f}	156.03±25.99 ^{c,e,f}	0.000*
Body Mass Index (kg/m ²)	33.16±1.28 ^{a,b,c}	37.32±1.48 ^{a,d,e}	43.70±2.65 ^{b,d,f}	56.88±6.60 ^{c,e,f}	0.000*
Lean Body Mass (kg)	37.21±13.96 ^{b,c}	39.29±15.82 ^{d,e}	47.20±2.65 ^{b,d,f}	59.29±22.31 ^{c,e,f}	0.000*
Skeletal Muscle Mass (kg)	29.91±6.98 ^{b,c}	31.21±6.14 ^e	32.93±5.82 ^{b,f}	39.18±6.99 ^{c,e,f}	0.000*
Body fat (%)	45.62±5.28 ^b	48.60±5.17	53.86±31.70 ^b	55.67±2.54	0.009*
Absolute Body Fat (kg)	51.17±12.98 ^{a,b,c}	60.91±13.31 ^{a,d,e}	71.93±15.97 ^{b,d,f}	96.74±19.28 ^{c,e,f}	0.000*
Lean Mass/Body Fat Ratio (kg)	0.82±0.43	0.71±0.37	0.71±0.29	0.64±0.24	0.052
Neck Circumference (cm)	37.91±3.91 ^{b,c}	39.01±4.06 ^e	40.73±4.33 ^{b,f}	43.46±4.92 ^{c,e,f}	0.000*
Waist Circumference (cm)	96.27±7.56 ^{a,b,c}	103.86±9.78 ^{a,d,e}	114.24±10.05 ^{b,d,f}	130.67±16.72 ^{c,e,f}	0.000*
Abdomen Circumference (cm)	105.40±7.62 ^{a,b,c}	114.62±10.52 ^{a,d,e}	126.81±10.83 ^{b,d,f}	147.63±15.32 ^{c,e,f}	0.000*
Hip Circumference (cm)	117.03±10.56 ^{a,b,c}	124.37±12.75 ^{a,d,e}	131.80±11.06 ^{b,d,f}	148.41±16.21 ^{c,e,f}	0.000*
Waist/Height Ratio (cm)	0.59±0.04 ^{a,b,c}	0.63±0.05 ^{a,d,e}	0.69±0.05 ^{b,d,f}	0.79±0.09 ^{c,e,f}	0.000*
Hemodynamic					
Variables/Physical Fitness Related to Health					
Diastolic Blood Pressure (mmHg)	123.25±14.70 ^c	127.32±14.52	127.41±13.78	132.25±11.79 ^c	0.027*
Systolic Blood Pressure (mmHg)	79.67±12.08	81.65±12.54	82.43±10.27	84.85±12.49	0.178
SPO ₂ (%)	96.71±3.36	96.51±12.54	96.04±2.35	92.26±1.77	0.197
HR (bpm)	77.80±8.89 ^b	80.76±12.87	84.06±11.26 ^b	84.89±12.66	0.001*
Six Minutes' Walk Test (m)	505.33±86.25 ^c	496.02±73.95 ^e	485.63±70.27 ^f	431.83±81.54 ^{c,e,f}	0.000*
Plank Strength Test (s)	28.88±26.85	27.96±24.54	25.31±22.81	17.05±14.55	0.116
Dynamic Lower Limb Muscular Endurance (n rep.)	15.72±4.54	15.16±4.69	14.40±3.78	13.78±4.29	0.064
Flexibility (cm)	22.79±8.14 ^{b,c}	19.62±9.86 ^d	14.67±7.77 ^{b,d}	15.14±10.27 ^c	0.000*
Biochemical Parameters					
Glycemia (mg/dL)	95.25±12.16 ^b	101.73±30.87	111.96±50.52 ^b	106.70±31.43	0.010*
Insulin (mU/L)	18.68±9.15 ^c	23.02±11.39	22.35±10.99	28.52±14.89 ^c	0.001*
Homa IR	4.45±2.46 ^{b,c}	5.73±3.13	6.22±4.42 ^b	7.25±3.60 ^c	0.001*
Homa β	67.34±32.87 ^c	81.68±44.17	74.49±38.71 ^f	99.86±58.88 ^{c,f}	0.002*
US-CRP (mg/L)	4.02±3.44 ^{b,c}	5.81±5.35	7.52±6.50 ^b	8.45±5.43 ^c	0.000*
Total cholesterol (mg/dL)	192.74±40.01	190±36.17	196.22±38.30	179.78±38.65	0.161
HDL-c (mg/dL)	49.92±12.25	46.73±11.99	46.74±12.36	48.78±15.78	0.236
LDL-c (mg/dL)	117.08±36.94	113.84±30.74	119.05±31.51	107.47±30.33	0.259
VLDL-c (mg/dL)	23.99±11.34	27.88±15.19	28.68±15.79	23.16±8.41	0.051
Non-HDL Cholesterol (mg/dL)	140.34±39.93	141.78±36.03	149.40±36.76	135.93±37.63	0.12
Triglycerides (mg/dL)	127.53±65.03	145.55±84.86	158.25±106.17	126.15±74.53	0.061
Glycated Hemoglobin (%)	5.52±0.54	5.66±0.98	5.79±1.44	5.50±0.79	0.288
Indices Derived From Biochemical/Anthropometric Parameters					
AIP (mg/dL)	2.83±1.88	3.44±2.63	3.85±3.41	3.03±3.14	0.067
MetS - Z BMI	0.38±0.55	0.86±0.79	1.35±1.24	1.67±0.66	0.059
Percentile BMI	63.35±18.91 ^{a,b,c}	75.34±15.43 ^{a,d,e}	83.70±13.35 ^{b,d,f}	92.52±6.84 ^{c,e,f}	0.000*
MetS-Z WC	0.17±0.60 ^{a,b,c}	0.60±0.83 ^{a,d}	1.03±1.17 ^{b,d}	1.10±0.67 ^c	0.000*
Percentile WC	55.43±21.08 ^{a,b,c}	67.39±18.57 ^{a,d,e}	75.83±17.73 ^{b,d}	82.30±13.05 ^{c,e}	0.000*
TYG (mg/dL)	8.59±0.53 ^b	8.74±0.58	8.87±0.71 ^b	8.65±0.57	0.011*
TYG-BMI	284.92±21.22 ^{a,b,c}	326.44±25.10 ^{a,d,e}	387.97±42.01 ^{b,d,f}	493.35±75.64 ^{c,e,f}	0.000*
TYG-WC	828.17±94.33 ^{a,b,c}	909.97±117.98 ^{a,d,e}	1015.51±137.87 ^{b,d,f}	1133.68±179.47 ^{c,e,f}	0.000*

^aObesity I vs. Obesity II; ^bObesity I vs. Obesity III; ^cObesity I vs. Super Obesity; ^dObesity II vs. Obesity III; ^eObesity II vs. Super Obesity; ^fObesity III vs. Super Obesity; One Way Anova test and Bonferroni post-hoc were used for differences; SPO₂- oxygen saturation; HR-heart rate; bpm: beats per minute; mg/dl – milligrams per deciliter; us-CRP – ultra-sensitive C-reactive protein; HDL – high density lipoprotein; LDL – low density lipoprotein; AIP- atherogenic index of plasma; BMI- Body Mass Index; TYG -Triglyceride-Glucose Index; WC- Waist Circumference. Data are shown as mean values \pm standard deviations.

The biochemistry parameters indicated significant differences in glucose, insulin, HOMA-IR, Homa-beta, and hs-PCR among the class III and IV compared to class I and II with the higher BMI classes presenting the unhealthier results ($p < 0.01$). However, the markers of dyslipidemia and the HbA1c levels did not show significant differences. Lastly, related to the index or ratios applied to identify alterations related to insulin resistance (IR) or dyslipidemia, significant differences were observed in the percentile of the continuous metabolic syndrome score related to BMI, with the higher classes of obesity presenting the more severe score risks of metabolic syndrome ($p < 0.001$). The same pattern is observed in the MetS-WC and Percentile of MetS-WC. The TYG parameters also have that pattern with the groups with obesity classes III and IV presenting the higher risks ($p < 0.001$).

When we look to the data stratified by sex (Tables 2 and 3), we noticed that among males (Table 2) there were significant differences among BMI groups except for age, height and fat percentage in the group of anthropometric and body composition variables, with higher values corresponding to the higher classes of BMI, with the only exception to the lean/fat ratio which has lower values in the higher BMI groups. Considering the male results for hemodynamic and physical fitness, the only variables presenting significant differences were heart rate (bpm), with higher values in the obesity class IV and lowers values to the distance in the six minutes' walk test and in the sit and reach (flexibility) test for the higher BMI groups. With the biochemistry parameters, we observed significant differences only with the high sensitive C-reactive protein (hsCRP), with higher values in the class IV BMI and in total cholesterol, LDL-c and non-HDL-c with, contrary to what was expected, lower values in the class IV BMI. Whereas in the last group of variables (biochemistry index or ratios), the significant differences were observed for MetS - Z BMI, and its respective percentile, and also for the MetS – Z WC and its percentile, always with higher values, indicative of the severity of the MetS in the class IV BMI group. There were also significant differences in the TYG-BMI and TYG-WC following the same pattern for the highest groups of BMI (classes III and IV).

Table 2. Cardiometabolic risk variables stratified by age group among a sample of Brazilian adults (n=404).

Variables	Young Adults (n=197)	Middle Age Adults (n=207)	p-value
Age (years)	31.84 \pm 5.26	46.32 \pm 5.17	0.000*
Height (m)	166.16 \pm 7.36	162.57 \pm 9.12	0.000*
Body Mass (kg)	114.90 \pm 22.65	103.78 \pm 20.57	0.000*
Body Mass Index (kg/m ²)	41.61 \pm 6.98	39.19 \pm 5.79	0.000*
Lean Body Mass (kg)	45.19 \pm 17.70	41.63 \pm 17.20	0.041*
Skeletal Muscle Mass (kg)	32.68 \pm 6.30	31.69 \pm 6.84	0.131
Body fat (%)	52.38 \pm 28.83	48.89 \pm 5.57	0.088
Absolute Body Fat (kg)	69.71 \pm 20.22	62.15 \pm 16.06	0.000*
Lean Mass/Body Fat Ratio (kg)	0.71 \pm 0.33	0.73 \pm 0.36	0.613
Neck Circumference (cm)	39.88 \pm 4.34	39.42 \pm 4.46	0.296
Waist Circumference (cm)	109.79 \pm 13.46	106.94 \pm 13.42	0.034*
Abdomen Circumference (cm)	122.42 \pm 14.55	117.53 \pm 15.44	0.001*
Hip Circumference (cm)	129.92 \pm 13.87	125.92 \pm 14.45	0.019*
Waist/Height Ratio (cm)	0.66 \pm 0.07	0.65 \pm 0.07	0.679
Hemodynamic Variables/Physical Fitness Related to Health			
Diastolic Blood Pressure (mmHg)	124.46 \pm 12.41	129.26 \pm 15.43	0.001*
Systolic Blood Pressure (mmHg)	80.89 \pm 10.36	82.65 \pm 12.66	0.128

SPO ₂ (%)	96.55±2.13	96.15±2.75	0.102
HR (bpm)	82.87±12.36	80.75±11.41	0.075
Six Minutes' Walk Test (m)	495.51±77.21	483.51±76.81	0.118
Plank Strength Test (s)	24.91±21.61	27.72±25.84	0.237
Dynamic Lower Limb Muscular Endurance (n rep.)	14.61±4.16	15.12±4.45	0.239
Flexibility (cm)	18.46±8.82	17.47±9.77	0.289
Biochemical Parameters			
Glycemia (mg/dL)	99.86±27.56	109.67±45.89	0.010*
Insulin (mU/L)	23.31±11.03	21.34±11.50	0.081
Homa IR	5.79±3.38	5.77±3.97	0.966
Homa β	88.22±41.80	71.75±41.33	0.006*
US-CRP (mg/L)	6.58±6.1	6.08±5.33	0.376
Total cholesterol (mg/dL)	184.65±35.05	199.56±39.37	0.000*
HDL-c (mg/dL)	45.53±12.22	49.31±12.47	0.002*
LDL-c (mg/dL)	110.99±29.17	120.93±34.33	0.002*
VLDL-c (mg/dL)	26.20±14.03	28.04±14.96	0.202
Non-HDL Cholesterol (mg/dL)	138.40±34.11	149.58±39.43	0.003*
Triglycerides (mg/dL)	143.02±98.22	148.62±83.27	0.536
Glycated Hemoglobin (%)	5.40±0.74	5.92±1.34	0.000*
Indices Derived From Biochemical/Anthropometric Parameters			
AIP (mg/dL)	3.58±3.28	3.34±2.48	0.41
MetS - Z BMI	0.97±0.86	1.06±1.17	0.267
Percentile BMI	77.34±17.53	77.73±16.71	0.818
MetS-Z WC	0.65±0.82	0.78±1.12	0.168
Percentile WC	68.82±19.99	70.10±20.19	0.524
TYG (mg/dL)	4.68±0.32	4.76±0.31	0.012*
TYG-BMI	361.83±70.31	346.85±59.86	0.021*
TYG-WC	955.34±154.02	947.52±152.18	0.608

t test for independent samples; SPO₂- oxygen saturation; HR-heart rate; bpm: beats per minute; mg/dl – milligrams per deciliter; us-CRP – ultra-sensitive C-reactive protein; HDL – high density lipoprotein; LDL – low density lipoprotein; AIP- atherogenic index of plasma; BMI- Body Mass Index; TYG -Triglyceride-Glucose Index; WC- Waist Circumference. Data are shown as mean values ± standard deviations.

Table 3. Proportion of participants presenting alteration in single parameters and biochemical index.

Single Parameter	Men (n=85)	Women (n=319)
Glycated Hemoglobin (%)	34.1	31.3
Non-HDL Cholesterol (mg/dL)	36.5	26.3
HDL-c (mg/dL)	40	45.1
LDL-c (mg/dL)	41.2	25.7
Insulin (mU/L)*	45.8	32.9
Triglycerides (mg/dL)	51.7	28.5
Total cholesterol (mg/dL)	52.9	49.5
Diastolic Blood Pressure (mmHg)	54.1	47.6
Glycemia (mg/dL)	57.6	36.9
Systolic Blood Pressure (mmHg)	77.6	59.8
Insulin (mU/L)**	84.7	83.1
US-CRP (mg/L)	89.4	89.9
Waist Circumference (cm)	90.6	95.9
Index or ratios	Man (n=85)	Women (n=319)
Homa IR	89.4	86.2

Homa β	95.3	94.4
MetS - Z BMI	95.3	92.1
AIP (mg/dL)	100	92.8

HDL – high density lipoprotein; LDL – low density lipoprotein; Insulin (mU/L)* - Egídio et al, 2017 ;us-CRP – ultra-sensitive C-reactive; Insulin (mU/L)**- Mcauley et, 2001; AIP- atherogenic index of plasma; BMI- Body Mass Index. Data are presented as proportion (%).

Comparing the BMI groups among the women (Table 3), we can see very similar results to those described above (Tables 1 and 2). In the anthropometric and body composition variables, only age and the ratio lean to fat mass didn’t show significant differences. Whereas all the other variables presented higher values for the classes III and IV of BMI. Considering the group of hemodynamic and physical fitness variables, there were significant differences among the BMI groups for heart rate, with the class I obesity presenting the lowest values. For the distance in the six minutes’ walk test, the sit and stand (dynamic) test and for the sit and reach (flexibility) test, we found that the higher BMI groups presented the lowest values. For the biochemistry parameters, there were significant differences in insulin, HOMA-IR, Homa-Beta, and hsCRP, all with higher values corresponding to the classes III and IV of BMI. The VLDL-c presented significant differences between the BMI class I against class II. Finally, for the biochemistry index or ratios, the significant differences were observed for MetS - Z BMI percentile and for the MetS – Z WC and its percentile, always with higher values, indicative of the severity of the MS in the class IV BMI group. There were also significant differences in the TYG 1 and 2 and TYG-BMI and TYG-WC following the same pattern with the highest values for groups of BMI (classes III and IV).

In Table 4, the participants are grouped according to their age group. They were split in two groups, the young adults (18 and 39 years old) and the middle age group with participants aged 40 to 50 years old. These groups were compared according to the same categories presented above with the BMI stratification. The differences here (Table 4) were observed in height, weight, BMI, LBM, FM, WC, abdominal circumference and hip circumference, all with higher values for the younger group. The other groups of variables presented different patterns, with the younger group having lower values for SBP and lower glucose levels. They also have total cholesterol, HDL-c, LDL-c, non-HDL-c, HbA1c, TyG1,2, and TyG-BMI lower than the middle age group. Whereas their HOMA-Beta was higher than the middle age group.

Table 5 and Figures 1 and 2 present the prevalence of single biochemical or anthropometric parameters and also indexes or ratios used to the assessment of metabolic risks. The lower prevalences were observed for HbA1c, with 34.1% of man and 31.3% of women showing elevated values (above of the cutoff point of 5.7%) for that parameter [26]. On the other hand, the index whose prevalence was higher was the AIP with 100% of man and 92.8% of women presenting values above the cutoff point of 0.1–0.24 as moderate risk, and >0.24 as high-risk occurrence of cardiovascular disease proposed by [27].

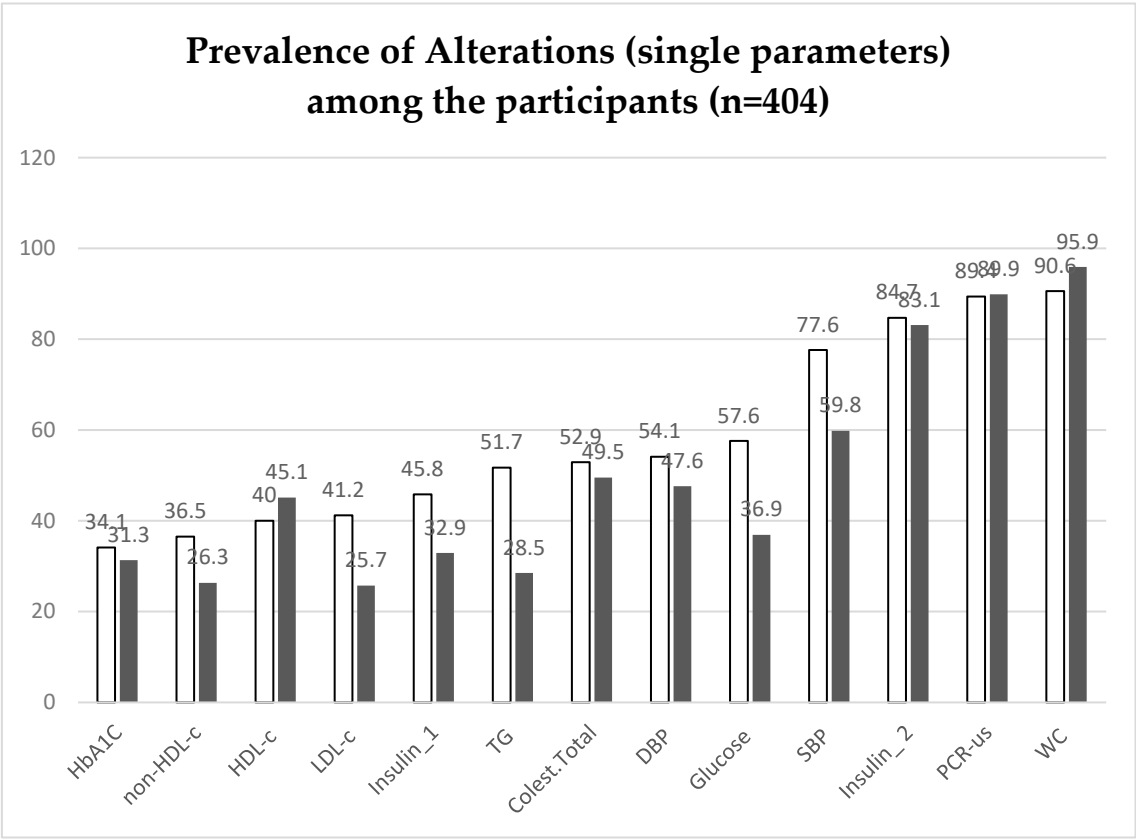


Figure 1. Prevalence of alterations in single parameters used in the diagnose of metabolic risk.

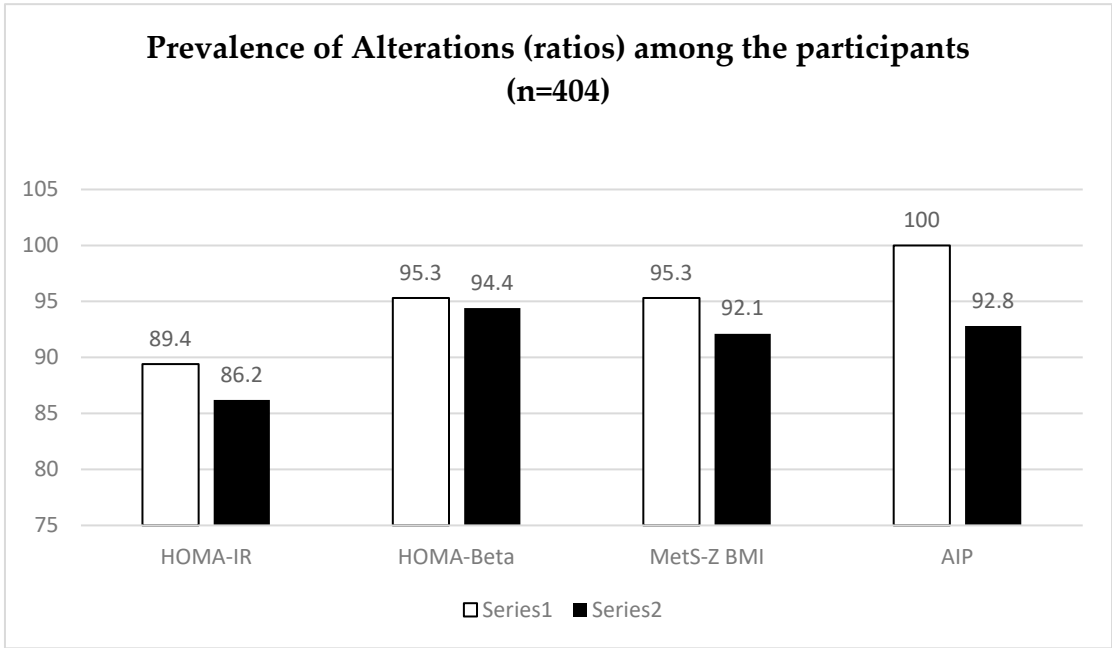


Figure 2. Prevalence of alterations in ratios or index based on the combination of single parameters used in the diagnose of metabolic risk.

It can be noted in Figure 3 that the score of severity of metabolic syndrome (MetSs) and the atherogenic index of plasma (AIP) presented a significant correlation with a coefficient of determination (R^2) of 69%, indicating that the two index are strongly correlated.

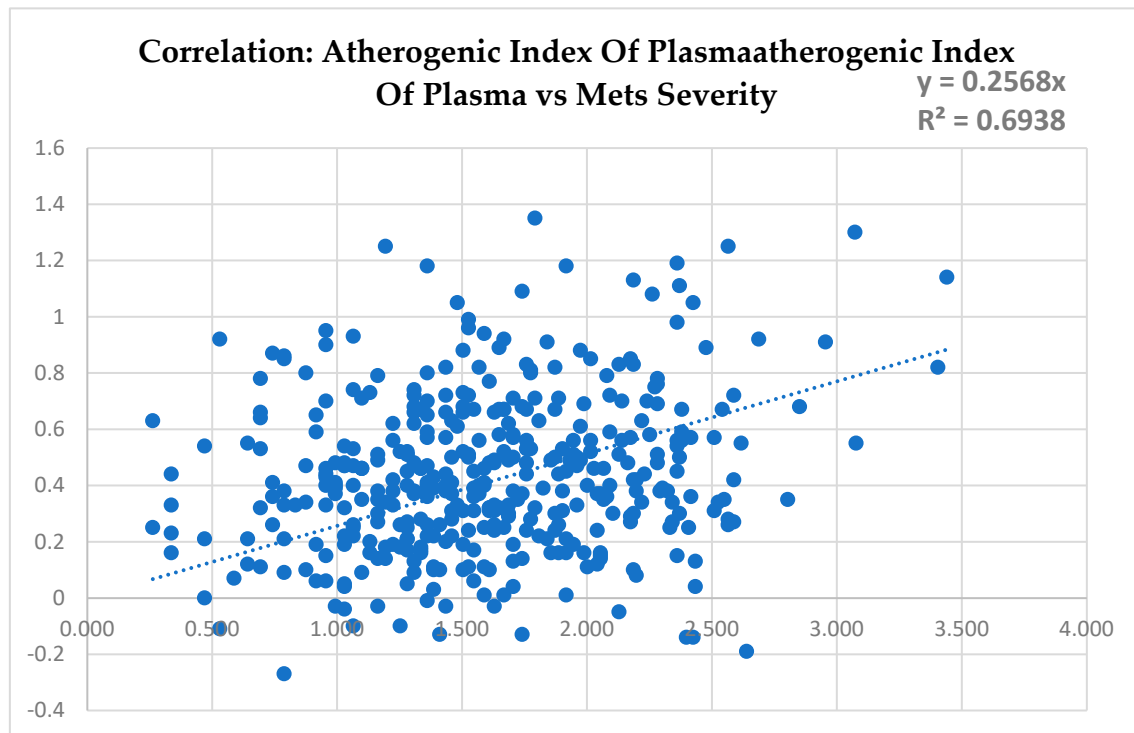


Figure 3. Relationship between two index of metabolic risk (the atherogenic index of plasma (AIP) and the score of severity of metabolic syndrome (MetS)).

4. Discussion

The main goal of this study was to verify the risk profile of participants in one multiprofessional treatment program of obesity (MTPO) offered regularly by the Maringa State University to different age groups since 2005. The adults involved in that program had a real need of professional care to treat their obesity and comorbidities. In Maringa, there is no public funded program offering that kind of treatment to this public. Therefore, this population needs to volunteer as research subjects to get access to that model of assistance. This is the first, and maybe the most important data to present. This reality is not different from the one observed in most of the countries around the world which also don't have enough MTPO to assist their population [8].

The second important information provided by this study is the high prevalence of metabolic risk found in this sample as even 100% of the male participants were classified as having moderate or high risk of occurrence of cardiovascular disease according to the AIP. The atherogenic lipoprotein profile of plasma has been recognized as a substantial risk factor for atherosclerosis. That is due to the significance of triglycerides in atherosclerotic and cardiovascular disease. In addition to individual serum cholesterol levels, the atherogenic index of plasma (AIP) has been suggested as a marker of plasma atherogenicity based on the evidence of its positive association with lipoprotein particle size, cholesterol esterification rates, and remnant lipoproteinemia [28].

The AIP was found to be one of the strongest markers in predicting the cardiovascular disease (CVD) risk. There is also evidence indicating that AIP is associated with other CVD risk factors. Thus, MTPO which promotes modification of lifestyle, is strongly recommended and the monitoring of parameters like AIP can be used to assess the efficacy or effectiveness of that kind of intervention program [29].

It is worth to explain that the AIP is defined as the logarithm of triglycerides to high-density lipoprotein cholesterol (HDL-C) ratio. Thus, it is a strong predictor of future cardiovascular disease. AIP was directly and independently associated with arterial stiffness and it is also known to inversely correlate with LDL particle size. Besides of that it can be readily calculated from the routine lipid profiles [21].

There is great interest in the development of new and comprehensive lipid index, like the atherogenic index of plasma, which might reflect the balance between atherogenic and anti-atherogenic factors. Recently, AIP has been shown to be a strong marker for predicting the risk of CAD and the value of AIP was positively associated with waist circumference and BMI and inversely associated with physical activity [30].

Another recent study found evidence to propose that AIP can be considered as a novel and better biomarker for obesity since subjects in the higher quartiles of AIP had all a significantly increased risk of obesity compared with those in the lowest quartile in a Chinese population [31]. Interestingly, the mean value for the group with obesity in that study was 0.13, whereas in our study that mean was 0.44 or more than three times higher. We also identified a similar fact comparing the results from our study to another applying the AIP in a Chinese population in which the average AIP was 0.17 to the group with coronary artery disease and 0.12 in the control group [30].

Therefore, results like these can be very important to balance the idea of metabolic health obese (MHO) which can mislead and postpone some more the intervention program that a higher number of individuals need. Besides, even when the MHO phenotype is correctly identified there is still a chance of reversing it over time, since the MHO is an instable state with an important part of the subjects evolving to the unhealthy phenotype within some years [9,11].

It is important to highlight too that even simple measures like the waist circumference can be very useful as a clinical tool, since that was the single parameter that diagnosed more than 90% in this study, regardless of gender. That inexpensive variable, alone or combined with more complex models, can be very useful for the risk stratification. But it is also important to express that increased WC is undeniably an excellent marker of cardiometabolic status in individuals with normal weight or overweight, but it is less useful in individuals with higher BMI [3].

In that context it becomes relevant to remind that the abdominally obese males increases the risk of cardiovascular disease 20-fold over the course of 5 years and that reinforce the necessity of protocols including simple measures like the WC as important tools to identify high risk individuals [7,32]. The same authors have also demonstrated that the excess fat occurs predominantly due to subcutaneous or visceral abdominal fat and are closely related to insulin resistance (IR) or diabetes.

The data presented in our study corroborated that with prevalence of HOMA-IR above the cutoff point over 80% in both genders with men presenting the higher prevalence (89.4%) with really close numbers from the women (86.2%). This is an impressive number since insulin resistance is one of the main factors associated with cardiovascular diseases. It is also important because in Brazil that diagnosis test is not included in the routine exams and therefore, an important proportion of persons with pre-diabetes or diabetes are not identified at the proper time [33].

It is also important to bring information about the HOMA-Beta utility since it is proposed as a measure of the functionality of the pancreatic beta cells. It has a cutoff point of 167 to 175 and it is interpreted as lower values indicating better beta cells functioning. Based on that cutoff point it was shown that a very high proportion of the studied population exposed to higher risk of long term damage of the beta cells functioning and, consequently, higher risks of the development of DM2 [18].

Another single parameter which shows very higher prevalence was the hsCRP with number close to 90% above the cutoff point. That reinforce the proinflammatory state promoted by the excess of adipose tissue which is common among people with obesity. This data is align with another study in which the hsCRP was the most important biomarker associated with the cardiovascular risk [34]. We have also a systematic review in which the hsCRP was used to stratify risks in several chronic non communicable diseases and it was recognized as the most promisor of the seric biomarkers to be employed in these conditions to assess the clinical status and evolution [35].

Despite the overall importance of the traditional way to diagnose the metabolic syndrome (MetS) there are difficulties related to the utility of this information for the present or future (follow up) use, due to its nature binary. That has been recognized and other ways to use it has been presented, like the continuous metabolic syndrome risk score (cMSy) as a more appropriate and valid alternative for epidemiological and clinical studies [12,13,36]. We applied the concept of the MetS severity score or MetSs to assess as a continuous variable with the mean is to represent one score of

risk, or severity of MetS. This score can be interpreted as a Z-score (mean 0, SD=1), with higher scores corresponding to a higher risk of MetS [14,36]. These authors have been revealing racial/ethnic discrepancies using the traditional MetS criteria. They have also found that some racial groups with diabetes had a low prevalence of MetS but they have high MetS severity scores with is a contradictory result. The same group also have found that the MetS severity score correlated with risk of future type 2 diabetes and CVD [37]. More importantly, maybe is the fact that their study showed no meaningful differences between the MetS-Z- WC and MetS-Z-BMI scores in their associations with future CHD and T2DM, indicating the potential utility for clinical use of MetS-Z-BMI.

That kind of tool has great potential to monitor prospectively high-risk patients since it has revealed that compared with individuals with a change in score smaller than (<0), participants with a change greater than half a point (>0.5) had a hazard ratio (HR) of 2.66 for incident diabetes ($p<0.001$). Our data show an increasing mean of both MetS-Z- WC and MetS-Z-BMI scores among the BMI classes as the higher BMI classes presenting significative higher averages, especially related to MetS-Z- WC.

Overall, these results made it clear that adults with obesity have higher risks to chronic non communicable diseases. They also have shown that different criteria, even the simple ones like WC can be very important tools to be used in the assessment of the population and they can have their diagnose improved combining results with more sophisticated parameters like those ratios and indexes presented in this study.

5. Conclusions

Finally, it is important to recognize that this study has some limitations like the cross-sectional design of it which does not allow conclusions about causality, therefore these findings should be confirmed in follow-up studies. On the other hand, it made it possible to verify that this population can be facing very challenging situations reflected by the high risks profile presented that have been ignored probably because of the fact that obesity has been recognized as a neglected disease around the world.

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