Research Article

Circulating Inflammatory Markers May Mediate the Relationship between Healthy Plant-Based Diet and Metabolic Phenotype Obesity in Women: A Cross-Sectional Study

Azam Mohamadi,1 Farideh Shiraseb,1 Atieh Mirzababaei,1 Dorsa Hosseininasab,2 Niloufar Rasaei,1 Cain C. T. Clark,3 and Khadijeh Mirzaei1

1Department of Community Nutrition, School of Nutritional Sciences and Dietetics, Tehran University of Medical Sciences (TUMS), Tehran, Iran
2Department of Nutrition, Science and Research Branch, Islamic Azad University, Tehran, Iran
3Centre for Intelligent Healthcare, Coventry University, Coventry, CV1 5FB, UK

Correspondence should be addressed to Khadijeh Mirzaei; mina_mirzaei101@yahoo.com

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1. Introduction

Obesity remains a leading public health concern all over the world and is associated with incidence of several major chronic diseases such as cardiovascular disease, type 2 diabetes, hypertension, dyslipidemia, and some types of cancer [1]. A systematic review and meta-analysis, in 2019, reported that the prevalence of obesity in Iranian older adults was 21.4% [2]; where there are criteria based on population- and country-specific cut-off point for Iranians to evaluate obesity, including body mass index (BMI) ≥ 30, waist circumference (WC) ≥ 91 cm for women, and ≥ 89 cm for men, and abdominal obesity, demarcated by increases in subcutaneous, deep subcutaneous, and intra-abdominal visceral adipose tissue [3–5]. In addition, individuals may be classified, metabolically, as having either healthy or unhealthy obesity on the basis of phenotypes.

Indeed, the metabolically healthy obesity (MHO) phenotype has been defined as favorable lipid profile as well as normal or slight changes in insulin sensitivity, while in metabolically unhealthy obesity (MUHO), these criteria are affected abnormally [6]. Some studies have indicated that an
intermediate-stage risk of metabolic disorders is evident in individuals with MHO in comparison with MUHO [7, 8].

It has been demonstrated that obesity and inflammation have a strong relationship, such that excessive fat mass can lead to an inflammatory response and increase markers of inflammation, such as transforming growth factor beta 1 (TGF-β1), interleukin-β1 (IL-β1), and monocyte chemoattractant protein-1 (MCP-1). The proinflammatory cytokines in obesity, emergent from the adipose tissue, are produced by infiltrating macrophages, as well as TGF-β1 having a role in the regulation of inflammation, immune function, and glucose homeostasis [9–11]. Thus, the adoption of a lifestyle is important to prevent chronic inflammation and other complications that might develop in obesity and chronic disease related to obesity.

According to cohort study, adopting a healthy plant-based diet may reduce the risk of cardiovascular disease in the general population, irrespective of genetic susceptibility [12]. Higher consumption of saturated fatty acid, salt, sugars, excessive alcohol, and low intake of fruit, vegetables, fiber, omega-3, and egg consumption are factors contributing to chronic disease [13, 14]. However, one review study amounts of eggs consumed by adults have no significant influence on systolic and diastolic blood pressure [15]. To reduce the inflammation factors and chronic diseases, higher intake of healthy plant foods (vegetable and fruit), instead of unhealthy foods (refined grains, sweets, and desserts), is useful [16, 17]. A review study shown that nut consumption especially walnut is contributing to weight reduction and control weight by reduction of fat absorption and appetite [18]. In a randomized controlled trial, significant improvement in weight and BMI was observed in the intervention group which had plant-based diet compared to the group with normal care [19]. Moreover, Kahleova et al., also observed a significant association between a plant-based diet with changes in body weight and body composition [20]. Empirical research has demonstrated that dietary patterns may be associated with inflammatory mediators and obesity as an inflammatory-related disease [21–24]. For instance, the positive impacts of dietary patterns, including Mediterranean, Dietary Approaches to Stop Hypertension (DASH), and plant-based diet (PBD) on inflammation, obesity, MHO, and MUHO have been elucidated by several studies [25–28]. Moreover, a review article suggested that PBD could reduce some inflammatory markers, such as IL-6 and C-reactive protein (CRP), but not TNF-α [26]. This type of dietary pattern has been defined as having high intake of fruits, vegetables, grains, and legumes, all of which could lead to reductions in both inflammation and obesity [29]. However, to our knowledge, the effect of PBD on MHO and MUHO has not yet been examined. Furthermore, dietary inflammatory index, in which the consumption of food items can increase body inflammation such as saturated fatty acid, trans fatty acids, and refined grains, has been shown to be positively associated with fat mass and concentration of MCP-1 [30]. Thus, adherence to a PBD may be associated with an improvement in obesity-related inflammatory profiles and prevention of chronic disease risks. To the best of our knowledge, this is the first study designed to assess the relationship between a healthy plant-based diet and metabolic obesity phenotype, in addition to investigating the mediating role of inflammatory markers (TGF-β1, IL-β1, MCP1) in overweight and obese Iranian women. Thus, we sought to investigate the influence of plant-based diet on MHO and MUHO phenotypes which are mediated by inflammatory markers in overweight and obese women.

2. Methods and Materials

2.1. Study Population. This cross-sectional study was conducted using multistage simple random sampling and participants consisted of 289 women recruited from 20 Tehran Health Centers in 2018. Indeed, 20 health centers were randomly selected from all health centers of the Tehran University of medical sciences. Sampling was such that people who referred to Tehran health centers, if they met the inclusion criteria, were randomly selected to enter the study. Inclusion criteria were age ≥18 years old, with a body mass index (BMI) ≥25 kg/m², without history of hypertension, had no intake of alcohol and opiate drugs, not being pregnant, not being in menopause, not having acute or chronic infection, and exclusion criteria were having history of cardiovascular disease, thyroid, cancer, diabetes, liver, and kidney disease, and smoking. In addition, participants who had been following any arbitrary special dietary regimen, as well as those with chronic disease(s) affecting their diet, or if their daily energy intake was <800 kcal or >4200 kcal [31], were excluded. All participants were asked to provide written informed consent prior to participation, and the study was approved by the ethics committee of Tehran University of Medical Sciences (IR.TUMS.VCR.REC.1395.1234). We calculate the study sample prior to the study. The sample size was computed according to the following formula:

\[ n = \left\lfloor \left( Z_1 - \alpha + Z_1 - \beta \right) \times \sqrt{1 - r^2} \right\rfloor r + 2 \right) = 289, \]  

where \( \beta = 0.95 \) and \( \alpha = 0.05 \), then, with 95% confidence and 80% power, and \( r = 0.37 \).

2.2. Anthropometric Measurements. Body composition, including weight, fat, and lean mass, and waist-to-hip ratio was assessed using a bioelectric impedance analyzer (In Body 770 scanner, Korea) [32]. Also, height was measured to the nearest 0.5 cm by nonelastic tape, while BMI was calculated as weight (kg) divided by height (m²). WC measurement was performed at the level of the umbilicus after exhalation. According to the World Health Organization (WHO) criteria for classification of weight, BMI ≥25 kg/m² was considered as overweight, and ≥30 kg/m² as obesity [33].

2.3. Biochemistry Measurements. Blood samples were drawn after 12 hours of overnight fasting to assess low-density lipoprotein (LDL-C), high-density lipoprotein (HDL-C), triglycerides (TG), homeostatic model assessment (HOMA), C-reactive protein (CRP), TGF-β1, IL-β1, and MCP-1, using ELISA. The serum was separated and stored at a temperature of −70°C until the analyses were carried.
2.4. Metabolic Health Phenotypes. Kareli criterion was used to estimate metabolic health, considering the following: TG ≤ 1.7 mmol/L or use of lipid-lowering drugs, HDL-C ≥ 1.3 mmol/L, LDL-C ≤ 2.6 mmol/L, HOMA ≤ 2.7, and CRP ≤ 3.0 mg/L [34]. Subsequently, participants were categorized into two groups as metabolically healthy obesity and unhealthy obesity phenotypes [35, 36], where meeting three or more of the preceding components equated to an unhealthy phenotype.

2.5. Dietary Intake Measurement. Dietary intake was collected using a 147-item semiquantitative food frequency questionnaire (FFQ), where its validity and reliability were previously affirmed in an Iranian population [37]. The average consumption frequency was considered as the past year on a daily, weekly, and monthly basis. Household measures were taken into account for portion sizes and then converted to grams [38]. The food composition table (FCT) of the US Department of Agriculture (USDA) was used to evaluate energy and nutrients. The Iranian FCT was considered for local foods that were not present in the USDA FCT. Moreover, total daily energy intake was examined by considering the sum of each food item energy.

2.6. Plant-Based Diet Scores. Plant foods were categorized into healthy and unhealthy groups based on epidemiological knowledge concerning the relationship between food items with obesity and inflammation [26]. First, we generated 18 food groups consisting of animal foods (butter/lard, dairy, egg, fish/seafood, meat, and miscellaneous animal-based foods), healthy (whole grains, fruits, vegetables, nuts, legumes, vegetable oils, and tea/coffee), and unhealthy plant foods (fruit juices, SSBs, refined grains, potatoes, and sweets/desserts), in line with nutrient and standard culinary features. Then, daily values and the number of servings for entire foods incorporated in each of the 18 food groups were summed. In this study, we generated an overall plant-based diet index (PDI) based on the algorithm developed by Martinez-Gonzalez et al. [39], and as suggested by Satija et al. two fitted versions of a healthful plant-based diet index (hpPDI), and an unhealthful plant-based diet index (upPDI) [40]. The cut points for the tertiles were calculated for each 18 food groups with an assigned score between 1 and 3 for each tertile. To achieve each participant’s indices, we summed the 18 food group scores. The observed score ranges for PDI, hpPDI, and upPDI were <51, 51–57, and >57, respectively, across the groups. It is worth mentioning that a higher amount of all indices is indicative of a lower intake of animal foods. Moreover, alcoholic beverages were not included in our indices due to their diverse association with several health outcomes.

2.7. Physical Activity Measurement. The international physical activity questionnaire (IPAQ), validated in Iranian women adults, was used to assess physical activity levels [41]. The participants were asked to answer questions such as the time they spent on walking, moderate, and vigorous physical activity during the last week. Then, the time of each physical activity was converted to minutes per week and calculated as metabolic equivalent of task (MET/minutes/week).

2.8. Other Covariates Assessments. Demographic characteristics including age, sex, income, marital status, supplement consumption, socioeconomic status, education, and occupation status were collected. In addition, systolic and diastolic blood pressure (SBP and DBP) were evaluated after 15-min rest, using a Mercury sphygmomanometer.

2.9. Statistical Analysis. Participants were categorized according to tertiles of PDI and hpPDI scores (tertiles 1: < 51, tertiles 2: 51–57, and tertiles 3: >57) and upPDI score (tertiles 1: <45, tertiles 2: 45–51, and tertiles 3: >51). Kolmogorov—Smirnov test and histogram were used to determine the normal distribution of the data. All variables with normal distribution were analyzed by parametric tests. One-way analysis of variance (ANOVA) for continuous variables and chi-square analysis for categorical variables were used to compare subject characteristics and dietary intake across tertiles of plant-based diet score and were reported as mean (SD). Analysis of covariance (ANCOVA) was used to examine demographic characteristics, anthropometric measurements, clinical assessments, and dietary intake across tertiles of PDI, hpPDI, and upPDI score, adjusting for age, BMI, physical activity, energy intake, occupation status, economic status, supplement consumption, and income. To examine the association between plant-based diet score and MHO and MUHO, binary logistic regression was used, and reported as odds ratio (OR) and 95% confidence interval (CI). Moreover, linear regression was used to examine the association of MHO and MUHO across the tertiles of PDI, hpPDI, and upPDI scores and reported as β and CI, and adjusted in model 2, including age, BMI, physical activity, energy intake; model 3, with occupation, economic status, supplement consumption, income; and using the Barrett method. Mediation analysis was performed to assess the mediating effects of TGF-β1, IL-β1, and MCP1 in models 4, 5, and 6, respectively. In the current study, SPSS software version 26 (Chicago—United State) was used for data analysis, and a P-value <0.05 was, a priori, considered statistically significant.

3. Results

3.1. Study Population Characteristics. In total, 289 women, including 65 MHO (26.9%) and 177 MUHO individuals (73.1%), with a mean age and BMI of 36.5 years old and 31.05 kg/m², respectively, were recruited in the present study. The mean (SD) height and weight of participants were 161.26 (5.92) cm and 80.70 (12.24) kg, respectively.

3.2. General Characteristics of Participants across the Tertiles of PDI, hpPDI, and upPDI Scores. General characteristics of participants, across the tertiles of PDI, hpPDI, and upPDI scores in overweight and obese women are shown in Table 1.
| Abbreviations: BMI, body mass index; DBP, diastolic blood pressure; FBS, fasting blood sugar; HDL, high-density lipoprotein; hPDI, healthful plant-based diet index; IL-β, interleukin-β; LDL, low-density lipoprotein; MCP1, monocyte chemoattractant protein-1; PDI, overall plant-based diet index; SBP, systolic blood pressure; TGF-β, transforming growth factor beta 1; WHR, waist-hip ratio; uPDI, unhealthful plant-based diet index. Values are mean ± standard deviation (SD) for continuous variables and percentage for dichotomous variables. Using one-way ANOVA for continuous variables and Chi-square test for categorical variables. *P-value adjusted for age, BMI, physical activity, energy intake. P-value <0.05 significant. n= Significant mean difference between tertiles one and two. n= Significant mean difference between tertiles one and three. |
Participants with a higher score of hPDI were more physically active \((P = 0.001)\), and those with a higher score of uPDI used less supplements (26.7%, \(P = 0.03\)). Also, individuals with a higher score of uPDI had a lower BMI \((P = 0.02)\) and WHR \((P = 0.02)\), after controlling for age, BMI, physical activity, and energy intake. The hs-CRP level was significantly higher across higher scores of PDI after adjusting confounders \((P = 0.02)\). However, no significant mean differences were found in IL-\(\beta_1\), MCP1, and TGF-\(\beta_1\) across PDI, hPDI, and uPDI.

### 3.3. Dietary Intake of Participants across the Tertiles of PDI, hPDI, and uPDI Scores

Dietary intake of participants across the tertiles of PDI, hPDI, and uPDI scores in overweight and obese women are presented in Table 2. Participants in the highest tertiles of PDI had higher intake of polyunsaturated fatty acid (PUFA), linoleic acid, linolenic acid, calcium, zinc, vitamin C, folate, total fiber, whole grains, fruits, dairy, and legumes \((P < 0.001)\). In addition, higher adherence of hPDI was associated with lower intake of PUFA, linoleic acid, linolenic acid, potassium, iron, magnesium, selenium, vitamin C, folate, total fiber, fruits, vegetables, nuts, legumes, and sea foods \((P < 0.001)\). Furthermore, women with the highest score of uPDI consumed higher cholesterol, refined grain, and lower SFA, MUFA, potassium, calcium, magnesium, zinc, vitamin A, \(\beta\) carotene, vitamin D, thiamin, vitamin B6, vitamin B12, whole grain, vegetables, dairy, legumes, sea foods, and animal oils \((P < 0.001)\) (Table 2). However, other dietary factors across tertiles of PDI, hPDI, and uPDI showed no significant results.

### 3.4. Association of Inflammatory Markers with PDI, hPDI, and uPDI Scores

Dietary intake of participants across the tertiles of PDI, hPDI, and uPDI scores in overweight and obese women are presented in Table 2. Participants in the highest tertiles of PDI had higher intake of polyunsaturated fatty acid (PUFA), linoleic acid, linolenic acid, calcium, zinc, vitamin C, folate, total fiber, whole grains, fruits, dairy, and legumes \((P < 0.001)\). In addition, higher adherence of hPDI was associated with lower intake of PUFA, linoleic acid, linolenic acid, potassium, iron, magnesium, selenium, vitamin C, folate, total fiber, fruits, vegetables, nuts, legumes, and sea foods \((P < 0.001)\). Furthermore, women with the highest score of uPDI consumed higher cholesterol, refined grain, and lower SFA, MUFA, potassium, calcium, magnesium, zinc, vitamin A, \(\beta\) carotene, vitamin D, thiamin, vitamin B6, vitamin B12, whole grain, vegetables, dairy, legumes, sea foods, and animal oils \((P < 0.001)\) (Table 2). However, other dietary factors across tertiles of PDI, hPDI, and uPDI showed no significant results.

4. Discussion

In this cross-sectional study, we presented, for the first time, the relationship between a healthy plant-based diet and metabolic obesity phenotype, in addition to investigating the mediating role of inflammatory markers (TGF-\(\beta_1\), IL-\(\beta_1\), MCP1), in overweight and obese women.

In the present study, hPDI and uPDI had no significant association with MHO and MUHO in the crude model, but after adjusting confounders, we noticed that women with higher adherence to hPDI, had a lower risk of MUHO phenotype. Kouvari et al. demonstrated that higher adherence to a plant-based diet was associated with a greater probability of long-term maintenance of a healthy metabolic state. In addition, the healthful or unhealthful food choices within this pattern appeared to strongly predict cardiometabolic condition in women [42].

In our study, we investigated the potential mediatory role of TGF-\(\beta_1\) in the association between healthy plant-based diet and metabolic phenotype obesity. Indeed, after...
Table 2: Dietary intake of participants across the tertiles of PDI, hPDI, and uPDI scores in overweight and obese women \( (n = 289) \).

| Nutrient                  | Mean (n = 289) | T2 (n = 111) | T3 (n = 105) | T1 (n = 79) | P-value |
|---------------------------|----------------|--------------|--------------|--------------|---------|
| Protein (g/d)             | 88.4 ± 28.50   | 90.35 ± 26.36| 89.49 ± 26.42| 96.53 ± 25.43| <0.001  |
| Carbohydrate (g/d)        | 372.59 ± 120.40| 302.72 ± 89.16| 374.57 ± 104.66| 458.64 ± 114.31| <0.001  |
| Total fat (g/d)           | 49.8 ± 33.78   | 80.39 ± 27.04| 98.2 ± 34.16 | 108.37 ± 33.28| 0.166   |
| Cholesterol (mg/d)        | 252.67 ± 104.94| 246.74 ± 93.31| 270.38 ± 124.37| 242.02 ± 94.45| <0.001  |
| SFA (g/d)                 | 28.02 ± 11.13  | 24.96 ± 9.47 | 26.7 ± 12.49 | 30.19 ± 10.84| 0.009   |
| MUFA (g/d)                | 31.32 ± 11.88  | 27.01 ± 10.26| 32.74 ± 11.76| 35.29 ± 12.27| 0.300   |
| PUFA (g/d)                | 20.5 ± 9.12    | 16.15 ± 6.94 | 20.78 ± 9.03 | 24.23 ± 9.68 | 0.048   |
| Linoleic acid (g/d)       | 17.33 ± 8.65   | 13.70 ± 6.60 | 18.09 ± 8.66 | 21.13 ± 9.17 | 0.002   |
| Linolenic acid (g/d)      | 1.23 ± 0.67    | 0.96 ± 0.46  | 1.25 ± 0.67  | 1.54 ± 0.76  | 0.047   |
| Trans fatty acid (g/d)    | 0.00 ± 0.00    | 0.00 ± 0.00  | 0.00 ± 0.00  | 0.00 ± 0.00  | 0.001   |
| Omega-6 fatty acid (g/d)  | 244.77 ± 143.27| 379.79 ± 1273.35| 425.58 ± 1429.39| 479.04 ± 1424.97| 0.611   |
| Omega-3 fatty acid (g/d)  | 432.85 ± 1548.07| 3567.15 ± 1296.01| 4439.34 ± 1553.59| 5105.30 ± 1380.30| 0.360   |
| Intake (g/d)              | 457.76 ± 146.97| 380.76 ± 1187.66| 471.36 ± 147.30| 5409.99 ± 129.28| 0.176   |
| Mg (mg/d)                 | 18.62 ± 5.94   | 15.51 ± 4.85 | 18.91 ± 5.62 | 22.44 ± 5.40 | 0.020   |
| Calcium (mg/d)            | 1162.04 ± 418.20| 1053.36 ± 366.21| 1222.07 ± 474.29| 1237.73 ± 672.68| <0.001  |
| Vitamin A (RAE/d)         | 772.34 ± 407.42| 666.11 ± 320.70| 822.60 ± 513.13| 854.27 ± 355.34| 0.459   |
| Vitamin E (mg/d)          | 12.88 ± 4.19   | 11.14 ± 3.47 | 13.33 ± 4.53 | 16.42 ± 3.82 | 0.077   |
| Selenium (mg/d)           | 119.69 ± 42.62 | 106.27 ± 36.78| 122.05 ± 41.73| 134.42 ± 45.53| 0.056   |
| Vitamin C (mg/d)          | 195.72 ± 112.53| 152.72 ± 93.87| 187.41 ± 96.65| 260.20 ± 157.06| 0.009   |
| Vitamin D (RAE/d)         | 195.72 ± 112.53| 152.72 ± 93.87| 187.41 ± 96.65| 260.20 ± 157.06| 0.009   |
| Vitamin B1 (g/d)          | 4.05 ± 1.19    | 3.28 ± 0.94  | 4.05 ± 1.19  | 4.06 ± 1.19  | 0.001   |
| Vitamin B12 (g/d)         | 4.33 ± 2.39    | 4.71 ± 2.94  | 4.71 ± 2.94  | 4.06 ± 1.19  | 0.001   |
| Total fiber (g/d)         | 45.18 ± 18.80  | 36.07 ± 14.35| 45.42 ± 17.40| 56.41 ± 18.94| 0.001   |
| Animal oils (g/d)         | 6.57 ± 12.23   | 6.08 ± 11.99 | 7.18 ± 13.23 | 5.94 ± 12.67 | 0.165   |

Abbreviations: MUFA, mono unsaturated fatty acid; PUFA, poly saturated fatty acid; SFA, saturated fatty acid. Values are mean ± standard deviation (SD) for continuous variables and Chi-square test for categorical variables. \( \text{P-value adjusted for energy intake. P-value <0.05 significant.} \)
adjusting for potential confounders, TGF-β1 had a significant inverse association with hPDI, in addition, the hs-CRP level was significantly higher across scores of PDI after adjusting confounders. The findings of Pourreza et al.’s study also support our findings, where the authors asserted that PDI was significantly associated with TGF-β1 and hs-CRP, and a significant positive association between uPDI and hs-CRP [43]. In contemporary work, in a cross-sectional study containing 240 middle-aged women, hPDI was significantly associated with reduced inflammatory biomarkers compared to uPDI [16]. The previous studies were shown consumption of healthy plant foods (vegetable and fruit), instead of unhealthy foods (refined grains, sweets, and desserts), can reduce the inflammation factors and chronic diseases [16, 17].

In another study, Kim et al. indicated that individuals in the highest quintile of uPDI had greater odds of metabolic syndrome (MetS) than those in the lowest quintile. Higher uPDI score was associated with higher odds of hypertriglyceridemia in men, and abdominal obesity, high fasting glucose, and hypertriglyceridemia in women, which was consistent with our study [44]. Moreover, in the present study, higher scores of uPDI were significantly associated with higher BMI, WHR, body fat percentage, TG, and LDL-C. One randomized controlled trial, showed significant reduction in weight and BMI in the intervention group which consumed plant-based diet compare to the group with normal care [19]. Also, Kahleova et al. observed a significant association between a plant-based diet and body weight reduction and body composition changes [20]. While in Kim et al. the positive associations between unhealthy plant-based diets and the components of MetS, such as abdominal obesity, high fasting glucose, and hypertriglyceridemia, were observed only in women. Moreover, the median score of uPDI was moderately higher in women than men in the highest quintiles of uPDI, and the authors posited that women may have consumed more unhealthy plant foods, such as refined grains (e.g., white rice or noodles), than men [44].

There are various mechanisms through which plant-based diets may be associated with obesity and inflammation. Indeed, plant-based diets may reduce body fat through decreased caloric intake and increased energy expenditure due to increased thermogenesis. Polyphenols and unsaturated fatty acids can affect the liver, muscle, and adipose tissue, to help upregulate the expression of peroxisome proliferator-activated receptor (PPAR), which augments oxidation leading to a reduced circulating pool of free fatty acids (FFAs), thereby reducing the accessibility of FFA for adipose tissue uptake and hypertrophy. Reduced use of saturated fats, which are commonly derived from animal-based foods, may also contribute to better insulin sensitivity [45]. Plant-based foods are a major source of phytochemicals, which can act as ligands, substrates, inhibitors, and cofactors for multiple enzymes [46]. The use of phytochemicals, particularly polyphenols, which are available in various plant foods (e.g., berries, grapes, onions, apples, cacao, green tea, soy, and whole grains), are associated with decreased mortality and chronic disease risk [47–50]. Adopting a plant-based diet has been shown to reduce cardiovascular disease risk in the general population, regardless of genetic susceptibility [12]. Based on a review study, walnut consumption has been associated with weight loss and weight control by reducing fat absorption and appetite [18]. Polyphenols are hydroxylated bioactive compounds that can also affect body fat, and an inverse association between polyphenol utilization and body weight has been reported [51, 52]. The food compound of an unhealthy plant-based diet may have higher intakes of undesirable nutrients and lower intakes of micronutrients and antioxidants, which can unfavorably affect metabolic syndrome and its (co)factors. Moreover, a high intake of added sugar from unhealthy plant foods would negatively affect lipid metabolism, glucose control, and weight gain [53], while decreased dietary fiber may impact glycemic control, insulin sensitivity, and lead to increases inflammation. Indeed, these effects could be related to reduced inflammation and oxidative stress [54, 55]. Hence, plant-based diets may provide advantages in the inhibition of chronic disease

| Table 3: Association of inflammatory markers with PDI, hPDI, and uPDI scores in overweight and obese women (n = 289). |
|---------------------------------------------------------------|
| PDI \( \beta \) 95% CI | P-value | hPDI \( \beta \) 95% CI | P-value | uPDI \( \beta \) 95% CI | P-value |
|-----------------------|---------|---------------------|---------|---------------------|---------|
| TGF-β1                |         |                     |         |                     |         |
| Model 1               | -0.046 (-182.204, 96.348) | 0.544 | -0.128 (-250.583, 19.523) | 0.093 | 0.110 (-42.275, 271.783) | 0.151 |
| Model 2               | 0.014 (-180.109, 206.647) | 0.892 | -0.201 (-347.819, -22.570) | 0.026 | 0.121 (-59.211, 316.404) | 0.178 |
| Model 3               | 0.028 (-190.705, 246.097) | 0.802 | -0.286 (452.996, -85.258) | 0.004 | 0.161 (37.575, 384.440) | 0.106 |
| IL-β1                 |         |                     |         |                     |         |
| Model 1               | 0.044 (-0.024, 0.035) | 0.710 | 0.002 (-0.03, 0.03) | 0.989 | 0.015 (-0.035, 0.040) | 0.901 |
| Model 2               | 0.051 (-0.031, 0.043) | 0.738 | -0.075 (-0.047, 0.07) | 0.593 | 0.066 (-0.030, 0.049) | 0.628 |
| Model 3               | 0.090 (-0.030, 0.052) | 0.592 | 0.036 (-0.036, 0.045) | 0.823 | -0.044 (-0.049, 0.036) | 0.762 |
| MCP1                  |         |                     |         |                     |         |
| Model 1               | -0.062 (-2.712, 0.993) | 0.361 | -0.004 (-1.914, 1.791) | 0.948 | 0.008 (-1.98, 2.241) | 0.903 |
| Model 2               | -0.127 (-4.258, 0.669) | 0.152 | -0.094 (-3.351, 0.811) | 0.230 | 0.040 (-1.708, 2.952) | 0.599 |
| Model 3               | -0.087 (-4.145, 1.580) | 0.378 | -0.095 (-3.712, 1.038) | 0.268 | 0.115 (0.007, 1.949) | 0.059 |

Abbreviations: CI, confidence interval; IL-β1, interleukin-beta 1; MCP1, monocyte chemoattractant protein-1; TGF-β1, transforming growth factor beta 1. Linear regression was used. Model 1: crude Model 2: adjusted for age, BMI, physical activity, energy intake. Model 3: adjusted for model 2 further with occupation status, economic status, supplement consumption, income.
Table 4: Association of MHO and MUHO phenotypes across the tertiles of PDI, hPDI, and uPDI scores in overweight and obese women (n = 289).

|        | T1 ≤51 (n = 111) | T2 51–57 (n = 90) | T3 ≥57 (n = 88) | P-trend | P-value | T1 ≤51 (n = 97) | T2 51–57 (n = 105) | T3 ≥57 (n = 87) | P-trend | P-value | T1 ≤45 (n = 104) | T2 45–51 (n = 99) | T3 ≥51 (n = 86) | P-trend | P-value |
|--------|------------------|------------------|------------------|---------|---------|------------------|------------------|------------------|---------|---------|------------------|------------------|------------------|---------|---------|
| MUHO   |                  |                  |                  |         |         |                  |                  |                  |         |         |                  |                  |                  |         |         |
| Model 1| 1                | 0.770 (0.398–1.531) | 1.020 (0.479–2.216) | 0.427   | 0.614   | 1.290 (0.633–2.617) | 0.750 (0.383–1.467) | 0.539 | 0.720   | 1.290 (0.613–2.717) | 0.750 (0.383–1.467) | 0.539 | 0.320   |         |         |
| Model 2| 1                | 0.907 (0.350–2.290) | 1.400 (0.565–3.699) | 0.765   | 0.566   | 1.837 (0.701–4.648) | 0.936 (0.424–2.065) | 0.040 | 0.310   | 1.837 (0.701–4.648) | 0.936 (0.424–2.065) | 0.040 | 0.310   | 1.354 (0.682–2.687) | 1.806 (0.876–3.721) | 0.429 | 0.277   |
| Model 3| 1                | 0.697 (0.248–2.056) | 0.907 (0.353–2.498) | 0.475   | 0.719   | 1.754 (0.628–4.985) | 0.953 (0.396–2.302) | 0.038 | 0.430   | 1.754 (0.628–4.985) | 0.953 (0.396–2.302) | 0.038 | 0.430   | 0.820 (0.315–2.132) | 1.094 (0.425–2.644) | 0.674 | 0.421   |

Abbreviation: CI, confidence interval; MUHO, metabolically unhealthy obesity; OR, odds ratio. Reference group: Metabolic healthy obesity Binary logistic regression was used. Model 1: Crude Model 2: adjusted for age, BMI, physical activity, energy intake. Model 3: adjusted for model 2 further with occupation status, economic status, supplement consumption, income.
Table 5: Association of MHO and MUHO phenotypes across the tertiles of PDI, hPDI, and uPDI scores mediated by inflammatory markers in overweight and obese women (n = 289).

| Inflammatory markers | PDI | hPDI | uPDI |
|----------------------|-----|------|------|
|                      | T1 ≤51 (n=111) | T2 51–57 (n=80) | T3 ≥57 (n=88) | T1 ≤51 (n=105) | T2 51–57 (n=87) | T3 ≥57 (n=87) | T1 ≤45 (n=104) | T2 45–51 (n=99) | T3 ≥51 (n=86) |
| IL-β1                | 3.465 (0.348, 34.486) | 1.104 (0.137, 8.877) | 0.300 (0.503) | 8.743 (0.655, 116.769) | 0.934 (0.137, 8.877) | 0.140 (0.174) | 0.828 (0.115, 1.246) | 0.743 (0.124, 1.246) | 1.759 (1.246) |
| MCP-1                | 0.767 (0.767, 7.67) | 1.220 (1.220, 12.20) | 0.565 (0.631) | 0.785 (0.517, 2.343) | 0.631 (0.517, 2.343) | 0.156 (0.327) | 0.565 (0.214, 2.434) | 0.635 (0.214, 2.434) | 0.854 (0.214, 2.434) |
| TGF-β1               | 0.391 (0.391, 0.785) | 0.785 (0.785, 0.785) | 0.156 (0.327) | 1.276 (0.348, 4.156) | 1.276 (0.348, 4.156) | 0.565 (0.156) | 0.565 (0.156, 4.156) | 0.749 (0.156, 4.156) | 0.681 (0.156, 4.156) |

Abbreviation: IL-β1, interleukin beta-1; MCP1, monocyte chemoattractant protein; MUHO, metabolically unhealthy obesity; TGF-β1, transforming growth factor beta-1. Reference group: Metabolic healthy obesity, OR, odds ratio. Binary logistic regression was used. \(^5\)Adjusted with age, BMI, physical activity, energy intake, occupation status, economic status, supplement consumption, income, and IL-β1. \(^6\)Adjusted with age, BMI, physical activity, energy intake, occupation status, economic status, supplement consumption, income, and MCP-1. \(^7\)Adjusted with age, BMI, physical activity, energy intake, occupation status, economic status, supplement consumption, income, and TGF-β1.
beyond decreased fat mass. Systemic concentrations of proinflammatory mediators are known to be higher in obese (BMI 30 kg/m²) vs. normal-weight persons [56, 57]; indeed, the elevated abdominal fat mass is related to a chronic increase of the circulating concentrations of inflammatory mediators containing multiple acute-phase inflammatory proteins including CRP [58, 59]. It should be noted that the liver and the lymphoid organs are generally the main production sites of these inflammatory mediators, but in obesity, adipose tissue becomes the main producer, resulting in chronic and permanent local and systemic inflammatory [60].

The present study possesses numerous strengths and limitations. First, to the best of our knowledge, this is the first study to have evaluated the association between a healthy plant-based diet and metabolic phenotype obesity and also investigated the potential mediating role of inflammatory markers (TGF-β1, IL-β1, and MCP1) in overweight and obese Iranian women. Second, another strength of this study is the recruitment of a large sample of obese and overweight individuals. In addition, dietary intake was assessed using a locally validated questionnaire, the FFQ which was completed via interview with an experienced dietitian to minimize measurement errors. Nevertheless, despite these strengths, we must acknowledge some limitations in the present study. First, the cross-sectional nature of this study limited the ability to suggest a causal relationship between a healthy plant-based diet and metabolic phenotype obesity. Second, small errors may be present in the dietary assessment, mostly due to misremembering the data and misclassification errors. Third, because our study only included women, the results are not generalizable to men, although clearly this was not the aim of the study.

5. Conclusion

In conclusion, a higher hPDI score was associated with a lower MUHO phenotype in overweight and obese Iranian women, which could be mediated by TGF-β1, IL-β1, and MCP1. Based on these data, consumption of a plant-based diet containing unrefined and whole plant-foods may have beneficial health effects. It is vital to consider the quality of plant foods consumed in the general population for the improvement of health outcomes.

Abbreviations

- BMI: Body mass index
- DBP: Diastolic blood pressure
- FBS: Fasting blood sugar
- HDL: High density lipoprotein
- hPDI: Healthful plant-based diet index
- IL-β1: Interleukin-beta 1
- LDL: Low density lipoprotein
- MCP1: Monocyte chemoattractant protein-1
- PDI: Overall plant-based diet index
- SBP: Systolic blood pressure
- TGF-β1: Transforming growth factor beta 1
- WHR: Waist-hip ratio

- uPDI: Unhealthful plant-based diet index
- MUFA: Mono unsaturated fatty acid
- PUFA: Poly saturated fatty acid
- SFA: Saturated fatty acid
- CI: Confidence interval
- MUHO: Metabolically unhealthy obesity
- OR: Odds ratio
- MHO: Metabolic healthy obesity
- WC: Waist circumference
- PBD: Plant-based diet
- DASH: Dietary approaches to stop hypertension
- CRP: C-reactive protein
- WHO: World health organization
- HOMA: Homeostatic model assessment
- FFQ: Food frequency questionnaire
- FCT: Food composition table
- USDA: United states department of agriculture
- IPAQ: Physical activity questionnaire
- MetS: Metabolic syndrome
- PPAR: Peroxisome proliferator-activated receptor
- FFA: Free fatty acids

Data Availability

It is available if needed.

Ethical Approval

Each participant was informed completely regarding the study protocol and provided a written and informed consent form before taking part in the study. The study protocol was approved by the ethics committee of Tehran University of Medical Sciences (TUMS) with the following identification IR.TUMS.MEDICINE.REC.1400.710.

Consent

All authors approved the final manuscript and consent for publication.

Conflicts of Interest

The authors declare that they have no conflicts of interest.

Authors’ Contributions

Azam Mohamadi (AM1), Atieh Mirzababaei (AM2), Khadijeh Mirzaei (KhM) designed the search; AM1 and AM2 conducted the sampling; Farideh Shiraseb performed statistical analysis; AM1, AM2, Dorsa Hosseininasab, Niloufar Raseai, Cain Clark (CC), and KhM wrote the paper; Khadijeh Mirzaei holds primary responsibility for final content. All authors read and approved the final manuscript.

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References

[1] G. A. Bray, K. K. Kim, J. P. H. Wilding, and W. O. Federation, “Obesity: a chronic relapsing progressive disease process. a position statement of the world obesity federation,” Obesity Reviews, vol. 18, no. 7, pp. 715–723, 2017.

[2] A. Vaisi-Raygani, M. Mohammadi, R. Jalali, A. Ghobadi, and N. Salari, “The prevalence of obesity in older adults in Iran: a systematic review and meta-analysis,” BMC Geriatrics, vol. 19, no. 1, pp. 1–9, 2019.

[3] A. Delavari, M. H. Forouzanfar, S. Alikhani, A. Sharifian, and R. Kelishadi, “First nationwide study of the prevalence of the metabolic syndrome and optimal cutoff points of waist circumference in the middle east,” Diabetes Care, vol. 32, no. 6, pp. 1092–1097, 2009.

[4] A. Misra, J. S. Wasir, N. K. Vikram, R. Guleria, and P. Kumar, “Cutoffs of abdominal adipose tissue compartments as measured by magnetic resonance imaging for detection of cardiovascular risk factors in apparently healthy adult Indian men in North India,” Metabolic Syndrome and Related Disorders, vol. 8, no. 3, pp. 234–247, 2010.

[5] S. P. Bhatt, A. Misra, P. Nigam, R. Guleria, and M. A. Q. Pasha, “Phenotype, body composition, and prediction equations (Indian fatty liver index) for non-alcoholic fatty liver disease in non-diabetic Asian Indians: a case-control study,” PLoS One, vol. 10, no. 11, Article ID e0142260, 2015.

[6] A. Karelis, M. Brochu, and R. Rabasa-Lhoret, “Can we identify metabolically healthy but obese individuals (MHO)?” Diabetes and Metabolism, vol. 30, no. 6, pp. 569–572, 2004.

[7] C. K. Kramer, B. Zinman, and R. Retnakaran, “Are metabolically healthy overweight and obesity benign conditions?” Annals of Internal Medicine, vol. 159, no. 11, pp. 758–769, 2013.

[8] R. Caleyachetty, G. N. Thomas, K. A. Toulis et al., “Metabolically healthy obese and incident cardiovascular disease events among 3.5 million men and women,” Journal of the American College of Cardiology, vol. 70, no. 12, pp. 1429–1437, 2017.

[9] M. F. Gregor and G. S. Hotamisligil, “Inflammatory mechanisms in obesity,” Annual Review of Immunology, vol. 29, no. 1, pp. 415–445, 2011.

[10] C.-S. Kim, H.-S. Park, T. Kawada et al., “Circulating levels of MCP-1 and IL-8 are elevated in human obese subjects and associated with obesity-related parameters,” International Journal of Obesity, vol. 30, no. 9, pp. 1347–1355, 2006.

[11] H. Yadav, C. Quijano, A. K. Kamaraju et al., “Protection from obesity and diabetes by blockade of TGF-β/smad3 signaling,” Cell Metabolism, vol. 14, no. 1, pp. 67–79, 2011.

[12] Y. Heianza, T. Zhou, D. Sun, F. B. Hu, J. E. Manson, and L. Qi, “Genetic susceptibility, plant-based dietary patterns, and risk of cardiovascular disease,” American Journal of Clinical Nutrition, vol. 112, no. 1, pp. 220–228, 2020.

[13] A. Noce, A. Romanì, and R. Bernini, “Dietary intake and chronic disease prevention,” Nutrients, vol. 13, no. 4, p. 1358, 2021.

[14] S. M. S. Islam, T. D. Purnat, N. T. A. Phuong, U. Mwingira, K. Schacht, and G. Fröschl, “Non-communicable diseases (NCDs) in developing countries: a symposium report,” Globalization and Health, vol. 10, no. 1, pp. 1–8, 2014.

[15] R. Kolahdouz-Mohammadi, M. Malekahmadi, Z. S. Clayton et al., “Effect of egg consumption on blood pressure: a systematic review and meta-analysis of randomized clinical trials,” Current Hypertension Reports, vol. 22, no. 3, pp. 1–9, 2020.

[16] P. Bolori, L. Setayesh, N. Rasaee, F. Jarrahi, M. S. Yekaninejad, and K. Mirzaei, “Adherence to a healthy plant diet may reduce inflammatory factors in obese and overweight women—a cross-sectional study,” Diabetes & Metabolic Syndrome: Clinical Research Reviews, vol. 13, no. 4, pp. 2795–2802, 2019.

[17] H. Yarizadeh, L. Setayesh, N. Majidi et al., “Nutrient patterns and their relation to obesity and metabolic syndrome in Iranian overweight and obese adult women,” Eating and Weight Disorders-Studies on Norexia, vol. 27, no. 3, pp. 1–11, 2021.

[18] N. Pahlavani, D. Rostami, F. Ebrahimi, and F. Azizi-Soleiman, “Nuts effects in chronic disease and relationship between walnuts and satiety: review on the available evidence,” Obesity Medicine, vol. 17, Article ID 100173, 2020.

[19] N. Wright, L. Wilson, M. Smith, B. Duncan, and P. McHugh, “The BROAD study: a randomised controlled trial using a whole food plant-based diet in the community for obesity, ischaemic heart disease or diabetes,” Nutrition and Diabetes, vol. 7, no. 3, p. e256, 2017.

[20] H. Kahleova, J. McCann, J. Alwarith et al., “A plant-based diet in overweight adults in a 16-week randomized clinical trial: the role of dietary acid load,” Clinical Nutrition ESPEN, vol. 44, pp. 150–158, 2021.

[21] M. C. Calle and C. J. Andersen, “Assessment of dietary patterns represents a potential, yet variable, measure of inflammatory status: a review and update,” Disease markers, 2019, Article ID 3102870, 13 pages, 2019.

[22] L. Galland, “Diet and inflammation,” Nutrition in Clinical Practice, vol. 25, no. 6, pp. 634–640, 2010.

[23] J. A. Dias, E. Wirfält, I. Drake et al., “A high quality diet is associated with reduced systemic inflammation in middle-aged individuals,” Atherosclerosis, vol. 238, no. 1, pp. 38–44, 2015.

[24] J. Barbaresco, M. Koch, M. B. Schulze, and U. Nöthlings, “Dietary pattern analysis and biomarkers of low-grade inflammation: a systematic literature review,” Nutrition Reviews, vol. 71, no. 8, pp. 511–527, 2013.

[25] C. P. Rodríguez-López, M. C. Gonzalez-Torres, C. A. Aguilar-Salinas, and O. D. A. S. H. Nájera-Medina, “Diet as a proposal for improvement in cellular immunity and its association with metabolic parameters in persons with overweight and obesity,” Nutrients, vol. 13, no. 10, p. 3540, 2021.

[26] F. Eichelmann, L. Schwingshackl, V. Fedirko, and K. Aleksandrova, “Effect of plant-based diets on obesity-related inflammatory profiles: a systematic review and meta-analysis of intervention trials,” Obesity Reviews, vol. 17, no. 11, pp. 1067–1079, 2016.

[27] S. D’Innocenzo, C. Biagi, and M. Lanari, “Obesity and the Mediterranean diet: a review of evidence of the role and sustainability of the Mediterranean diet,” Nutrients, vol. 11, no. 6, p. 1306, 2019.

[28] H. Farhadnejad, M. Darand, F. Teymoori, G. Asghari, and A. Malekinejad, “Adherence to a healthy plant diet may reduce inflammatory factors in obese and overweight women—a cross-sectional study,” Diabetes & Metabolic Syndrome: Clinical Research Reviews, vol. 13, no. 4, pp. 2795–2802, 2019.
women: a cross-sectional study,” *BMC Research Notes*, vol. 13, no. 1, pp. 544–547, 2020.

[31] W. Willett, “Issues in analysis and presentation of dietary data,” *Nutritional Epidemiology*, vol. 15, no. 3, pp. 321–346, 1998.

[32] A. Mirzababaei, F. Shiraseb, F. Abaj et al., “The effect of dietary total antioxidant capacity (DTAC) and Caveolin-1 gene variant interaction on cardiovascular risk factors among overweight and obese women: a cross-sectional investigation,” *Clinical Nutrition*, vol. 40, no. 8, pp. 4893–4903, 2021.

[33] "WHO," pp. 180–181, 1970, https://www.euro.who.int/en/health-topics/disease-prevention/nutrition/a-healthy-lifestyle/body-mass-index-bmi.

[34] A. D. Karelis and R. Rabasa-Lhoret, “Inclusion of C-reactive protein in the identification of metabolically healthy but obese (MHO) individuals,” *Diabetes and Metabolism*, vol. 34, no. 2, pp. 183–184, 2008.

[35] A. Mirzababaei, S. F. Sajjadi, N. Ghodoosi et al., “Relations of major dietary patterns and metabolically unhealthy overweight/obesity phenotypes among Iranian women,” *Diabetes and Metabolic Syndrome: Clinical Research Reviews*, vol. 13, no. 1, pp. 322–331, 2019.

[36] A. Mirzababaei, K. Mirzaei, L. Khorrami-Nezhad, Z. Maghbooli, and S. A. Keshavarz, “Metabolically healthy/unhealthy components may modify bone mineral density in obese people,” *Archives of Osteoporosis*, vol. 12, no. 1, pp. 95–99, 2017.

[37] P. Mirmiran, F. H. Esfahani, Y. Mehrabi, M. Hedayati, and A. Mirzababaei, K. Mirzaei, L. Khorrami-Nezhad, H. Kim, K. Lee, C. M. Rebholz, and J. Kim, “Association of W. Willett, “Issues in analysis and presentation of dietary data,” *Nutritional Epidemiology*, vol. 15, no. 3, pp. 321–346, 1998.

[38] P. Mirmiran, F. H. Esfahani, Y. Mehrabi, M. Hedayati, and A. Mirzababaei, K. Mirzaei, L. Khorrami-Nezhad, H. Kim, K. Lee, C. M. Rebholz, and J. Kim, “Association of W. Willett, “Issues in analysis and presentation of dietary data,” *Nutritional Epidemiology*, vol. 15, no. 3, pp. 321–346, 1998.

[39] D. R. Bassett Jr., “Commentary to accompany,” *Public Health Nutrition*, vol. 40, no. 8, pp. 4893–4903, 2021.

[40] M. Ghaffarpour, A. Houshiar-Rad, and H. Kianfar, “The manual for household measures, cooking yields factors and edible portion of foods,” *Tehran: Nashre Olume Keshavarzy*.

[41] A. M. Martinez-González, A. Sanchez-Tainta, D. Corella et al., “A provegetarian food pattern and reduction in total mortality in the prevencion con dieta mediterranea (PREDIMED) study,” *American Journal of Clinical Nutrition*, vol. 100, no. suppl_1, pp. 320S–328S, 2014.

[42] A. Satija, S. N. Bhupathiraju, E. B. Rimm et al., “Plant-based dietary patterns and incidence of type 2 diabetes in US men and women: results from three prospective cohort studies,” *PLoS Medicine*, vol. 13, no. 6, Article ID e1002039, 2016.

[43] D. R. Bassett Jr., “Commentary to accompany,” *Medicine and Science in Sports and Exercise*, vol. 35, no. 8, p. 1396, 2003.

[44] M. Kouvari, D. B. Panagiotakos, C. Chrysohoou et al., “Healthful and unhealthful plant-based dietary patterns and their role on 10-year transition to metabolically unhealthy status in obese participants of the ATTICA prospective (2002–2012) study,” *European Heart Journal*, vol. 41, no. Supplement_2, 2020.

[45] S. Pourreza, Z. Khademi, A. Mirzababaei et al., “Association of plant-based diet index with inflammatory markers and sleep quality in overweight and obese female adults: a cross-sectional study,” *International Journal of Clinical Practice*, vol. 75, no. 9, Article ID e14429, 2021.

[46] H. Kim, K. Lee, C. M. Rebholz, and J. Kim, “Association between unhealthy plant-based diets and the metabolic syndrome in adult men and women: a population-based study in South Korea,” *British Journal of Nutrition*, vol. 125, no. 5, pp. 577–590, 2021.

[47] J. Rienks, J. Barbaresko, and U. Nöthlings, “Association of polyphenol biomarkers with cardiovascular disease and mortality risk: a systematic review and meta-analysis of observational studies,” *Nutrients*, vol. 9, no. 4, 2017.

[48] S. M. Nachvak, S. Moradi, J. Anjoom-Shoae et al., “Soy, soy isoflavones, and protein intake in relation to mortality from all causes, cancers, and cardiovascular diseases: a systematic review and dose-response meta-analysis of prospective cohort studies,” *Journal of the Academy of Nutrition and Dietetics*, vol. 119, no. 9, pp. 1483–1500, 2019.

[49] R. Kimble, K. M. Keane, J. K. Lodge, and G. Howatson, “Dietary intake of anthocyanins and risk of cardiovascular disease: a systematic review and meta-analysis of prospective cohort studies,” *Critical Reviews in Food Science and Nutrition*, vol. 59, no. 18, pp. 3032–3043, 2019.

[50] P. Knekt, J. Kumpulainen, R. Järvinen et al., “Flavonoid intake and risk of chronic diseases,” *The American Journal of Clinical Nutrition*, vol. 76, no. 3, pp. 560–568, 2002.

[51] M. L. Bertoia, E. B. Rimm, K. J. Mukamal, F. B. Hu, W. C. Willett, and A. Cassidy, “Dietary flavonoid intake and weight maintenance: three prospective cohorts of 124,086 US men and women followed for up to 24 years,” *BMJ*, vol. 352, p. i17, 2016.

[52] J. A. Vernarelli and J. D. Lambert, “Flavonoid intake is inversely associated with obesity and C-reactive protein, a marker for inflammation, in US adults,” *Nutrition and Diabetes*, vol. 7, no. 5, p. e276, 2017.

[53] K. L. Stanhope, “Sugar consumption, metabolic disease and obesity: the state of the controversy,” *Critical Reviews in Clinical Laboratory Sciences*, vol. 53, no. 1, pp. 52–67, 2016.

[54] R. S. Najjar, C. E. Moore, and B. D. Montgomery, “Consumption of a defined, plant-based diet reduces lipoprotein, inflammation, and other atherogenic lipoproteins and particles within 4 weeks,” *Clinical Cardiology*, vol. 41, no. 8, pp. 1062–1068, 2018.

[55] M. Romeu, N. Aranda, M. Giralt, B. Ribot, M. R. Nogues, and V. Arijas, “Diet, iron biomarkers and oxidative stress in a representative sample of Mediterranean population,” *Nutrition Journal*, vol. 12, no. 1, p. 102, 2013.

[56] C. Herder, M. Peltonen, W. Koenig et al., “Systemic immune mediators and lifestyle changes in the prevention of type 2 diabetes,” *Diabetes*, vol. 55, no. 8, pp. 2340–2346, 2006.

[57] C. Herder, T. Illig, W. Rathmann et al., “Inflammation and type 2 diabetes: results from KORA Augsburg,” *Gesundheitswesen*, vol. 67, no. Suppl 1, pp. S115–S121, 2005.

[58] M. Visser, L. M. Bouter, G. M. McQuillan, M. H. Wener, and T. B. Harris, “Elevated C-reactive protein levels in overweight and obese people,” *BMJ*, vol. 319, no. 7199, pp. 139–150, 2000.

[59] D. G. Cook, M. A. Mendall, P. H. Whincup et al., “C-reactive protein concentration in children: relationship to adiposity and other cardiovascular risk factors,” *Atherosclerosis*, vol. 149, no. 1, pp. 139–150, 2000.

[60] P. C. Calder, N. Ahluwalia, F. Brouns et al., “Dietary factors and low-grade inflammation in relation to overweight and obesity,” *British Journal of Nutrition*, vol. 106, no. Suppl 3, pp. S5–S78, 2011.