A dietary pattern rich in fruits and dairy products is inversely associated to gestational diabetes: a case-control study in Iran

Abazar Roustazadeh 1,2,†, Hamed Mir 3,4,†, Sima Jafarirad 3,5,*, Farideh Mogharab 2,6, Seyed Ahmad Hosseini 3,5, Amir Abdoli 2 and Saiedeh Erfanian 2,4

Abstract

Background: Gestational diabetes mellitus (GDM) causes many problems for mother and her neonate. A healthy diet plays an important role in preventing GDM. This study aimed to investigate the relationship between major dietary patterns and the GDM.

Methods: 386 healthy and 306 GDM pregnant women (total 693) completed this case-control study. Basic information and anthropometric indices were recorded, and a food frequency questionnaire was completed. For extracting major dietary patterns, the principal component analysis was performed. Multivariable logistic regression models were used to examine whether specific dietary patterns are associated to the GDM.

Results: Four dietary patterns were identified: “fruits and dairy products”, “red meat and plant-based foods”, “snacks and high-fat foods” and “carbohydrate-rich foods”. Among these major extracted dietary patterns, “fruits and dairy products” showed an inverse association to the GDM (odds ratio adjusted for confounders: 0.50, confidence interval: 0.284–0.882, p-trend = 0.019, for highest vs. lowest quartile).

Conclusions: It seems using a healthy dietary pattern such as “fruits and dairy products” may decrease GDM risk.

Keywords: Pregnancy, Gestational diabetes, Dietary pattern, Body mass index

Background

Gestational diabetes mellitus (GDM) is a disorder during pregnancy, defined as glucose intolerance in women who have never had diabetes so far [1]. The GDM causes many complications during pregnancy or after childbirth for mother and her baby [2]. In 2017, it was reported 1 in 7 births was affected by the GDM worldwide [3]. GDM prevalence in Asia is estimated by 11.5% [4]. In Iran, a report showed a range from 1.3 to 18.6% in GDM prevalence [5]. However, GDM incidence in the population who live in urban area is similar to developed countries [6].

Obesity and a family history of diabetes are major risk factors for the GDM [7]. Obesity affects insulin resistance, so it may elevate GDM risk [8]. Having a diet with more energy intake from foods including sweet snacks and fats lead to overweight, obesity [9], or type 2 diabetes mellitus (T2DM) [10].

Previous studies focused on the relationship among nutrients or single foods and GDM [11, 12]. A dietary...
pattern shows the type of a diet and various foods which people habitually consume [13] which is related to the culture and social factors. It is a better index to assess the relationship between a disease and a diet in epidemiologic studies [14]. Some studies showed rich dietary patterns in the whole grains, fruits, vegetables, and nuts such as prudent and Mediterranean dietary patterns have an inverse relationship with the GDM [15, 16]. Besides, animal-based dietary patterns may increase GDM risk [17]. It seems a diet rich in fruit, vegetables, whole grains, and fish and low in red and processed meat, refined grains, and high-fat dairy can decrease in GDM risk [18].

There are some differences in the type of dietary patterns between Western and Asian countries due to different cultures. A study in China showed the relationship between the traditional Chinese diet and GDM risk [19]. It was an interesting finding, because it showed a traditional diet may lead to an increase in GDM risk. Although conflicting results showed there is no relationship between dietary patterns and GDM risk in the Asian countries [20].

Iran is an Asian country with various cultures and ethnicities. These different cultures may lead to various dietary patterns. There are limited studies about the relationship between dietary patterns and the GDM in Asia and Iran. There are conflicting results due to different cultures [19–21]. Therefore, we tried to find a relationship between major dietary patterns and the GDM in Jahrom, a city in the South of Iran.

**Methods**

**Participants**

This study is a case-control study conducted in Jahrom, a city in Fars province, South of Iran. Pregnant women who were referred to Motahari and Honari hospitals participated in the study. A simple sampling method was used for selecting subjects. The inclusion criteria for the case and control groups were pregnant women from 24th to 28th weeks of gestation, age ranges of 18–40 years, no history of diabetes or the GDM, having no chronic diseases (such as cardiovascular diseases, chronic renal disease, liver, and gastrointestinal diseases, hypo or hyperthyroidism, and severe anemia), singleton pregnancy and no weight loss program before pregnancy. Abortion, preeclampsia, eclampsia, and incomplete questionnaires were considered exclusion criteria in case and control groups. One-step procedure was used to diagnose the GDM in the case group. In this procedure, the oral glucose tolerance test was done. Blood glucose was measured after 8–12 h of fasting. Then, pregnant women were given 75 g of oral glucose, and blood glucose was measured again after one and two hours. The GDM was diagnosed if any blood glucose values were equal or more than 92 mg/dL, 180 mg/dL, and 153 mg/dL in fasting, 1 and 2 h after oral glucose consumption, respectively [22].

This study was a 1:1, case-control study. The required sample size was calculated assuming the relationship between major dietary patterns and the GDM, due to a study in Asia, which found a traditional dietary pattern was related to the odds of GDM (the highest quartile of traditional dietary pattern compared to the lowest quartile; odd ratio = 2.92) [19]. The sample size was calculated with Fleiss et al., method [23], using an online OpenEpi calculator (version 3) for case and control studies [24]. By considering 4% of controls exposed, the sample size was determined by 305 subjects in each case and control group (90% power of study and a 95% level of confidence). At first, 565 GDM and 634 healthy pregnant women were invited to study. 435 GDM and 493 healthy pregnant women accepted the invitation. Due to the inclusion criteria, 383 in the case and 454 in control have remained. 67 subjects in the case and 49 subjects in control refused participation in the study, and 721 pregnant women participated (316 in the case and 405 in control). Twenty-eight subjects (10 in the case and 18 in control) were excluded due to incomplete questionnaires. Finally, 693 subjects (306 in the case and 387 in control) were included in the analysis.

**General data collection**

Subjects completed a general questionnaire about their age, weight before pregnancy, number of children, active smoking before pregnancy (yes, no), level of education, and socioeconomic status (very low, lower medium, medium, high) [25]. The socioeconomic status was determined by asking about income. Then, the participants were divided into very low, lower medium, medium, and high. Level of education was categorized into illiterate or primary school, high school or diploma, college, and university. The level of physical activity was determined by the short form of international physical activity questionnaire [26]. The metabolic equivalent was determined which used to divide participants into sedentary, very low, low, medium, and hard activity levels. In the hospital, a trained technician with an accurate scale measured the weight of pregnant women monthly and recorded the weight and date of health card visit. Therefore, the health card was used to record the trend of weight gain during pregnancy. Pre-pregnancy body mass index (BMI) was calculated by reported pre-pregnancy weight in kilograms divided by the height in squared meters.

**Dietary assessment**

A food frequency questionnaire (FFQ) was used to estimate dietary intakes with 168 food items. The reliability
and validity of FFQ were shown in a study in the Iranian adult population which used twelve 24-hours dietary recalls to compare nutrients intake from FFQ [27]. This questionnaire was completed via a face-to-face interview with the trained nutritionists. Participants reported their intake frequency of an intended food in FFQ during last year, basis on daily, weekly, monthly, or annual frequency of intake. To increase dietary assessment accuracy, interviewers used a booklet which showed them the image of portion size. After completing FFQ, the frequency of each food item was converted into the daily intake. We used household measurements to convert each food to the gram. Besides, food intakes were adjusted for total energy intake by residual method [28]. Data were entered into SPSS (version 17.0, SPSS Inc., Chicago, IL, USA) software to extract dietary patterns.

| Table 1 Rotated factor loading in four major dietary patterns<sup>a</sup> | Fruits and dairy products | Red meat and plant-based foods | Snacks and high-fat foods | Carbohydrate-rich foods |
|---|---|---|---|---|
| 1 Fresh fruits (cantaloupe, melon, watermelon, pear, apricot, cherry, apple, peach, nectarine, plum, fig, grape, kiwi, grapefruit, orange) | 0.685 | 0.330 | |
| 2 Fruit Juices (grapefruit juice, orange juice, apple juice, cantaloupe juice, canned fruit) | 0.608 | |
| 3 High fat dairy products (high fat milk, cocoa milk, strained yogurt, high fat yogurt, traditional ice cream, other ice creams) | 0.602 | |
| 4 Low fat dairy products (low fat milk, skim milk, low fat yogurt, doogh (traditional yogurt drink)) | 0.554 | |
| 5 Creamy dairy products (creamy cheese, creamy yogurt, cream, butter) | 0.396 | |
| 6 Olive products (fresh olive, olive oil) | 0.370 | 0.325 | |
| 7 Raw vegetables (lettuce, tomato, cucumber, leek, parsley, scallion, basil, carrot, raw spinach, bell pepper) | 0.594 | |
| 8 Whole cereals (barley, oatmeal, corn (all types)) | 0.499 | |
| 9 Nuts and seeds (peanuts, walnuts, almonds, pistachios, hazelnuts, sunflower seeds) | 0.484 | |
| 10 Onion, garlic and cabbage (raw onion, garlic, pepper (red and black), cabbage (all types)) | 0.465 | |
| 11 Dried fruits (dried figs, raisins, dates, dried berries, dried apricots, dried peaches) | 0.451 | |
| 12 Red meats (beef, Lamb meat, minced lamb, sheep’s tongue, broth) | 0.406 | |
| 13 Legumes (lentil, pinto bean, pigeon pea, broad bean, mung bean, soybean) | 0.465 | 0.332 | |
| 14 Pickles and sauces (ketchup, traditional pickles, sour lemon juice) | 0.373 | 0.417 | |
| 15 Salt and salty snacks (salt, salty popcorn, chips) | 0.373 | 0.590 | |
| 16 Cakes and sweets (biscuit, cracker, tea cake, creamy cake, other cakes, sugar, tablet sugar, honey) | 0.587 | |
| 17 Organ Meats (tripe, sheep brain, leg of lamb, heart and liver of lamb, sausage) | 0.487 | |
| 18 Fats and oils (animal fat, mayonnaise, hydrogenated oil, cooking oils, margarine) | 0.466 | |
| 19 White meats (chicken, fish (all types), shrimp) | 0.447 | |
| 20 Traditional breads (Lavash, Barbari, Sangak, Taftoon) | 0.557 | |
| 21 Cooked vegetables (squash, stewed vegetables, eggplant, zucchini, celery; boiled carrot; fried onion; cooked spinach, mushroom, turnip) | 0.475 | |
| 22 Other cereals and starch sources (rice, pasta, potato, French fries, vermicelli, traditional vermicelli (Reshtie)) | 0.420 | |
| 23 White breads (baguette bread, toast, wheat flour) | -0.313 | |
| 24 Cheese and whey (cheese (all types), traditional whey (Kashk)) | |
| 25 Tea and coffee (tea (all types), coffee (all types)) | |
| Variance explained (%) | 11.93 | 6.67 | 5.81 | 5.03 |

<sup>a</sup>Factor loading < ± 0.3 was not shown
Dietary pattern extraction
Considering foods similarity, they were categorized into related food groups. Twenty-five food groups were included analysis (Table 1). Principal component analysis (PCA) was used to find major dietary patterns due to energy-adjusted foods intake. This analysis method is widely used which is an adaptive descriptive data analysis tool. Besides, it provides information regarding the maximum number of factors [29]. Kaiser-Meyer-Olkin (KMO) and Bartlett’s tests were used to assess factor analysis suitability. KMO ranges between 0 and 1, and a minimum value for good factor analysis is 0.6. Besides, Bartlett’s test should be significant ($p < 0.05$) [30]. The sampling sufficiency of components was approved by the KMO test > 0.67. Moreover, inter-correlation of components was confirmed using Bartlett’s test of sphericity < 0.001. We used a factor eigenvalue to decide several factors to retain. Besides, we used the scree plot which involves plotting each value. Dietary patterns were determined due to eigenvalue > 1.2 and scree plot examination (Fig. 1). Orthogonal rotation (varimax) was applied to simplify data interpretation. Factor loading more than 0.3 was chosen to find more relationships between food groups and dietary patterns. The dietary pattern was named using principal food groups in each pattern. A score was assigned to all participants considering adherence to each dietary pattern. Quartiles of factor scores were determined due to total participants and considered in the further analysis [31].

Statistical analysis
Kolmogorov-Smirnov tested the normality of data. Quantitative variables were compared between the case and control groups by independent sample $t$-test and categorical variables by the Chi-squared test. One-way analysis of variance was used to compare the mean of quantitative variables among quartile groups of the factor scores of dietary patterns. Tukey test was applied as a post-hoc analysis. Non-parametric tests were used in case of non-normal distribution. The logistic regression model was used to find the relation of independent variables and extracted dietary patterns with gestational diabetes. In this analysis, the lowest quartile was considered as a reference. Besides, this test was used to eliminate the effect of confounders. The odds ratio (OR) was determined with a 95% confidence interval (CI). Stratified analysis was applied to assess a relationship between the dietary pattern and the GDM stratified by pre-pregnancy BMI levels using Mantel-Haenszel test. All analyses were done using SPSS version 17.0 (version 17.0, SPSS Inc., Chicago, IL, USA). A $p$-value less than 0.05 was considered a significant level.

Results
This study was started from November 2018 which completed in September 2019. Total 721 pregnant women (405 healthy and 316 GDM) participated in the study. Twenty-eight subjects (18 in control and 10 in case)
were excluded because of incomplete FFQ. At last, 693 subjects (387 in control and 306 in case) were included in the analysis.

Table 2 shows the comparison between healthy and the GDM pregnant women for demographic, anthropometric, and socioeconomic variables. Subjects age with the GDM was significantly more than healthy \( (p < 0.001) \). Besides, pre-pregnancy BMI and weight gain during pregnancy in the GDM subjects were more than healthy ones \( (p < 0.001) \). There was no difference between two groups for smoking and job. Chi-square analysis showed a difference in socioeconomic status, physical activity, education, and the number of pregnancies between two case and control groups \( (p < 0.001) \).

Four major dietary patterns were extracted using PCA. Due to the scree plot, four dietary patterns with eigenvalues higher than 1.2 were selected as major dietary patterns (Fig. 1). These dietary patterns were: “fruits and dairy products”, “red meat and plant-based foods”, “snacks and high-fat foods,” and “carbohydrate-rich foods”. Table 1 shows the factor loading of food groups in each extracted dietary pattern. The cumulative variance of four dietary patterns was 29.45%. The “fruits and dairy products” pattern consisted of 11.93% of the variance and contained fresh fruits, fruit juices, olive, and dairy products. The “red meat and plant-based foods” pattern consisted of 6.67% of the variance and mainly contained vegetables, cereals, nuts, legume and red meats. The “snacks and high-fat foods” pattern consisted of 5.81% of the variance, and including cake and sweets, salty snacks (such as chips), organ meats, fats and oils, and white meats. The “carbohydrate-rich foods” pattern consisted of 5.03% of variance and mainly including traditional bread, other cereals, and starch sources (such as potato), and cooked vegetables.

Age, anthropometric indices, number of pregnancies, job, level of education, and socioeconomic status were compared among the quartiles of factor scores of dietary patterns (Table 3). Smoking and physical activity level were not shown because we did not find any significant value. Age of subjects was different among the quartiles of “fruits and dairy products”, also “snacks and high-fat foods” dietary patterns. In the fourth quartile of “snacks and high-fat foods,” pregnant women had a lower age than the first quartile. However, BMI of pre-pregnancy and weight gain during pregnancy were not different among each dietary pattern quartiles. There was a significant difference in education and the number of pregnancies among the quartiles of “fruits and dairy products” dietary pattern. Besides, there was a significant difference and a trend near significance for socioeconomic status and level of education among the quartiles of “fruits and dairy products” dietary pattern (Table 3).

Results showed “fruits and dairy products” pattern could inversely predict GDM \( (p \text{-trend} = 0.003) \). After adjustment for age, BMI of pre-pregnancy, weight gain during pregnancy, energy intake, socioeconomic status,
Because pre-pregnancy BMI may reflect some problems during pregnancy such as the GDM, we adjusted this effect by stratified analysis. Pre-pregnancy BMI was divided into four levels: underweight (< 18.5), normal (18.5–25), overweight (25–30), and obese (> 30). Due to the median of score, the score of factors of “fruits and dairy products” dietary pattern was divided into two levels. By adjusting pre-pregnancy BMI effect, the odds of GDM were lower in the higher median of “fruits and dairy products” dietary pattern (Mentel-Haenszel common OR: 0.580, CI: 0.413–0.810). It seems pre-pregnancy BMI is a confounding factor, but not interacting.

### Table 3 Comparison of age, anthropometric indices, number of pregnancies, and socioeconomic factors among quartiles (Q1-Q4) of major extracted dietary patterns

|                      | Fruits and dairy products | Red meat and plant-based foods | Snacks and high-fat foods | Carbohydrate-rich foods |
|----------------------|---------------------------|-------------------------------|---------------------------|-------------------------|
|                      | Q1    | Q4    | Q1    | Q4    | Q1    | Q4    | Q1    | Q4    | Q1    | Q4    |
| **Age**              |       |       |       |       |       |       |       |       |       |       |
|                      | 30.5 ± 5.7 | 29.2 ± 5.3 | 29.5 ± 5.9 | 30.2 ± 5.6 | 31.4 ± 5.4 | 28.3 ± 5.8 | 29.8 ± 5.4 | 30.0 ± 5.6 |
| p**                  | 0.010 | 0.255 | 0.019 | 0.001 | 0.018 | 0.001 | 0.019 | 0.001 |
| **Weight pre- pregnancy** | 66.4 ± 13.1 | 66.1 ± 11.6 | 64.8 ± 11.3 | 66.9 ± 10.6 | 67.2 ± 11.8 | 63.2 ± 13.1 | 66.9 ± 12.1 | 66.6 ± 13.6 |
| p**                  | 0.915 | 0.344 | 0.135 | 0.815 | 0.815 | 0.815 | 0.815 | 0.815 |
| **BMI pre- pregnancy** | 25.6 ± 4.8  | 25.8 ± 4.4  | 25.9 ± 4.4  | 25.8 ± 4.6  | 25.8 ± 4.4  | 26.4 ± 4.5  | 25.8 ± 4.2  | 26.4 ± 4.9  |
| p**                  | 0.996 | 0.844 | 0.178 | 0.235 | 0.235 | 0.235 | 0.235 | 0.235 |
| **Weight gain**      | 8.6 ± 5.1 | 8.1 ± 4.9 | 8.2 ± 5.1 | 8.0 ± 4.6 | 8.2 ± 4.6 | 8.2 ± 5.3 | 8.2 ± 5.2 | 8.1 ± 5.1 |
| p**                  | 0.120 | 0.768 | 0.997 | 0.530 | 0.530 | 0.530 | 0.530 | 0.530 |
| **Education**        |       |       |       |       |       |       |       |       |       |
|                      |       |       |       |       |       |       |       |       |       |
| Non or primary       | 30.6 % | 19.8 % | 25.4 % | 26 % | 18 % | 35.3 % | 26.7 % | 31.4 % |
| High school and diploma | 39.3 % | 43 % | 46.8 % | 38.7 % | 34.9 % | 42.2 % | 45.3 % | 37.8 % |
| College and university | 30.1 % | 37.2 % | 27.7 % | 35.3 % | 47.1 % | 22.5 % | 27.9 % | 30.8 % |
| p***                 | 0.061 | 0.086 | < 0.001 | 0.202 | 0.202 | 0.202 | 0.202 | 0.202 |
| **Number of pregnancies** |       |       |       |       |       |       |       |       |       |
|                      |       |       |       |       |       |       |       |       |       |
| First                | 33.1 % | 33.5 % | 29.7 % | 28.3 % | 24.3 % | 39.5 % | 32.9 % | 27.3 % |
| Second               | 32 % | 37.6 % | 32 % | 34.7 % | 37.6 % | 34.3 % | 30.6 % | 30.5 % |
| Third and more       | 34.9 % | 28.9 % | 38.4 % | 37 % | 38.2 % | 26.2 % | 36.4 % | 33.1 % |
| p***                 | 0.459 | 0.331 | 0.004 | 0.184 | 0.184 | 0.184 | 0.184 | 0.184 |
| **Job**              |       |       |       |       |       |       |       |       |       |
|                      |       |       |       |       |       |       |       |       |       |
| House wife           | 89 % | 87.2 % | 89.6 % | 88.4 % | 82.6 % | 91.3 % | 92.4 % | 88.3 % |
| Employee             | 8.1 % | 11.6 % | 6.9 % | 9.8 % | 15.1 % | 7.6 % | 5.2 % | 11.1 % |
| Self-employment      | 2.9 % | 12 % | 3.5 % | 1.7 % | 2.3 % | 1.2 % | 2.3 % | 0.6 % |
| p***                 | 0.533 | 0.357 | 0.134 | 0.003 | 0.003 | 0.003 | 0.003 | 0.003 |
| **Socioeconomic status** |       |       |       |       |       |       |       |       |       |
|                      |       |       |       |       |       |       |       |       |       |
| Very low             | 20.2 % | 6 % | 20.8 % | 10.8 % | 9.4 % | 17.3 % | 17.5 % | 13.5 % |
| Lower medium         | 27.2 % | 45.2 % | 27.2 % | 37.7 % | 38.8 % | 32.7 % | 33.7 % | 30.6 % |
| Medium               | 52.6 % | 47 % | 50.9 % | 49.1 % | 48.2 % | 48.8 % | 46.4 % | 54.7 % |
| High                 | 0 % | 1.8 % | 1.2 % | 2.4 % | 3.5 % | 1.2 % | 2.4 % | 1.2 % |
| p***                 | < 0.001 | 0.181 | 0.059 | 0.456 | 0.456 | 0.456 | 0.456 | 0.456 |

Data are presented as mean ± standard deviation for quantitative variables and percent for categorical variables

*One-way analysis of variances

**Non-normal distribution, Kruskal-Wallis test

***Chi-squared test
The relationship between the GDM and anthropometric indices and social factors was investigated. Besides, PCA was used to identify major dietary patterns and their association with the GDM.

"Fruits and dairy products" showed an inverse relationship with the GDM. Less prediction of GDM was seen in the third and fourth quartile of this pattern. Also, this inverse relationship was significant after adjustment for the effect of confounders. This pattern’s main foods were all kinds of dairy products, fresh fruits and fruit juices and all kinds of olive products. The GDM is characterized as glucose intolerance in pregnancy. It develops when a woman cannot produce enough insulin during pregnancy. Fruits and olives contain phytochemicals such as polyphenols. Polyphenols may affect glucose homeostasis by several mechanisms. These mechanisms include stimulating insulin secretion, controlling the digestion of carbohydrate and absorption of glucose, controlling the release of glucose from the liver and finally activating insulin receptors [32]. Therefore, dietary patterns, which are rich in fruits and olives, such as the Mediterranean diet, may help pregnant women to prevent the GDM. Although “fruits and dairy products” dietary pattern shows some differences from Mediterranean diet, it contains Mediterranean diet’s main foods, especially fruits and olives. Some documents showed that dairy products’ consumption is an effective factor in the prevention of T2DM [33]. A study investigated the effect of “dietary approaches to stop hypertension” (DASH) on insulin resistance in the GDM which showed the improvement of fasting blood glucose and serum insulin [34]. A DASH diet is due to whole grains, fruits, vegetables, and low-fat dairy products. Whereas, in “fruits and dairy products” pattern, there are two principal items of DASH diet. All kinds of dairy products were included in this dietary pattern. However, Anue et al. confirmed the protective effect of low-fat dairy products and cheese on T2DM incidence [33]. It seems the positive effect of dairy consumption is due to their potential calcium sources [35]. A study showed high calcium intake from diet is related to less incidence of the GDM [36]. Besides, dairy products contain whey protein as an important nutrient. Whey protein may improve hyperglycemia by several mechanisms including stimulation of insulin secretion and the action of incretin hormones such as glucagon-like polypeptide-1 and gastric inhibitory peptide [37].

“Snacks and high-fat foods” dietary pattern consists of more unhealthy foods. We did not find any relationship between this unhealthy dietary pattern and the GDM. However, the crude model was near to being significantly associated, but no significant value was observed after adjustment for confounders. The main foods in this dietary pattern included fats and oils, organ meats, white meats, legumes, sauces, cakes and sweets, salt and salty.

### Table 4

| Dietary pattern                  | Model 1 | Model 2 | Model 3 |
|---------------------------------|---------|---------|---------|
|                                 | N  | OR  | CI  | P-trend* | OR  | CI  | P-trend* | OR  | CI  | P-trend* |
| Fruits and dairy products       | Q1  | 175 | -   | 0.003    | 1  | -   | 0.004    | 1  | -   | 0.019    |
|                                 | Q2  | 176 | 0.623 | 0.383-1.013 | 0.668 | 0.399-1.20 | 0.575 | 0.327-1.011 |
|                                 | Q3  | 176 | 0.507 | 0.308-0.833 | 0.511 | 0.303-0.862 | 0.472 | 0.267-0.835 |
|                                 | Q4  | 172 | 0.478 | 0.29-0.785 | 0.488 | 0.289-0.825 | 0.500 | 0.284-0.882 |
| Red meat and plant-based foods  | Q1  | 177 | 1   | -   | 0.634    | 1  | -   | 0.798    | 1  | -   | 0.950    |
|                                 | Q2  | 169 | 0.964 | 0.591-1.571 | 0.869 | 0.519-1.455 | 0.901 | 0.512-1.585 |
|                                 | Q3  | 173 | 0.827 | 0.508-1.346 | 0.754 | 0.451-1.263 | 0.659 | 0.373-1.163 |
|                                 | Q4  | 174 | 0.742 | 0.456-1.209 | 0.712 | 0.422-1.200 | 0.660 | 0.372-1.173 |
| Snacks and high-fat foods       | Q1  | 175 | 1   | -   | 0.068    | 1  | -   | 0.750    | 1  | -   | 0.672    |
|                                 | Q2  | 184 | 1.794 | 1.108-2.905 | 1.889 | 1.139-3.135 | 1.709 | 0.988-2.957 |
|                                 | Q3  | 162 | 1.203 | 0.728-1.988 | 1.325 | 0.777-2.262 | 1.153 | 0.642-2.068 |
|                                 | Q4  | 172 | 0.800 | 0.484-1.321 | 1.082 | 0.617-1.897 | 0.931 | 0.498-1.741 |
| Carbohydrate-rich foods         | Q1  | 168 | 1   | -   | 0.569    | 1  | -   | 0.592    | 1  | -   | 0.559    |
|                                 | Q2  | 186 | 0.960 | 0.588-1.567 | 0.997 | 0.592-1.679 | 1.065 | 0.606-1.872 |
|                                 | Q3  | 167 | 1.408 | 0.852-2.329 | 1.542 | 0.905-2.629 | 1.743 | 0.970-3.130 |
|                                 | Q4  | 172 | 1.387 | 0.843-2.383 | 1.982 | 1.129-3.481 | 1.795 | 0.984-3.277 |

Model 1: crude model; Model 2: adjusted for age, the body mass index of pre-pregnancy, weight gain during pregnancy, and total energy intake; Model 3: additionally adjusted for socioeconomic status, level of education, physical activity and number of pregnancies

OR odds ratio, CI confidence interval

*The logistic regression model, the lowest quartile as reference
snakes which all of them except legumes were unhealthy. Although “snacks and high-fat foods” were similar to Western dietary pattern, legumes might neutralize unhealthy foods’ effect in this dietary pattern. In consistent to our study, two studies in Asia showed no relationship between Western dietary pattern and the GDM [20, 38]. However, two other studies found that using Western dietary pattern increased GDM risk [19, 21]. Besides, we did not find any relationship between “red meat and plant-based foods” and “carbohydrate-rich foods”, with GDM. There are some cultural factors, may affect eating habits which lead to different metabolic events [39]. Besides, diet composition heterogeneity may be another reason for no relationship among these dietary patterns with GDM [40].

Strength and limitation
The present study’s strength is an almost large sample size to extract the dietary patterns and homogeneity of subjects for ethnicity. This study has some limitations. Firstly, FFQ is a semi-quantitative questionnaire and depends on the food list which may cause some errors in food intake estimation, although we used a 168-food item food frequency questionnaire to extract dietary patterns which is a comprehensive questionnaire. Another limitation is using multiple questionnaires which made the mothers bored. Recall bias is also a limitation of study, because mothers were asked to report the food intake during the previous year. Besides, there is not a cause and effect association in a case-control study.

Iran contains different cultures, and it would be suggested for future research to study GDM risk concerning culture and dietary patterns in well-designed prospective studies.

Conclusions
Dietary pattern is an effective factor in the incidence of many chronic diseases. Finding an appropriate dietary pattern such as “fruits and dairy products” could help pregnant women to prevent the GDM. Further prospective studies and clinical trials are needed to confirm the results of this study.

Abbreviations
BMI: Body mass index; CI: Confidence interval; DASH: Dietary approaches to stop hypertension; FFQ: Food frequency questionnaire; GDM: Gestational diabetes mellitus; KMO test: Kaiser-Meyer-Olkin test; PCA: Principal component analysis; T2DM: Type 2 diabetes mellitus

Acknowledgements
The authors greatly appreciate pregnant women who participated in the study and Dr. Amirhossein Ramezani for his useful comments.

Authors’ contributions
SJ, AR and FM designed the protocol of the study. HM, AA and SE conducted the study. SJ analyzed the data and drafted the manuscript. SAH interpreted data and critically revised the manuscript. All authors read and approved the final manuscript.

Funding
This article is a part of the Ph.D. thesis of Hamed Mir, which approved in Nutrition and Metabolic Diseases Research Center, Ahvaz Jundishapur University of Medical Sciences (ID: NRC-97/03) and Jahrom University of Medical Sciences (ID: 2830).

Availability of data and materials
The datasets generated and analyzed during the current study are available from the corresponding author on reasonable request.

Declarations
Ethics approval and consent of participate
This study was approved by the Ethics committee of Ahvaz Jundishapur University of Medical Sciences (ID: IR.AJUMS.REC.1397.203) and Jahrom University of Medical Sciences (ID: IR.JUMS.REC.1396.148), based on the declaration of Helsinki. All participants completed a consent form and their information was kept secret.

Consent for publication
Not applicable.

Competing interests
The authors declare no competing interests.

Author details
1Department of Biochemistry, School of Medicine, Jahrom University of Medical Sciences, Jahrom, Iran. 2Department of Advanced Medical Sciences and Technologies, School of Medicine, Jahrom University of Medical Sciences, Jahrom, Iran. 3Nutrition and Metabolic Diseases Research Center, Clinical Research Institute, Ahvaz Jundishapur University of Medical Sciences, Ahvaz, Iran. 4Research Center for Non-communicable Diseases, Jahrom University of Medical Sciences, Jahrom, Iran. 5Department of Nutrition, School of Allied Medical Sciences, Ahvaz Jundishapur University of Medical Sciences, Ahvaz, Iran. 6Department of Obstetrics and Gynecology, Jahrom University of Medical Sciences, Jahrom, Iran.

Received: 30 April 2020 Accepted: 23 February 2021
Published online: 04 March 2021

References
1. Melchior H, Kurch-Bek D, Mund M. The Prevalence of Gestational Diabetes A Population-Based Analysis of a Nationwide Screening Program. Dtsch Arztebl Int. 2017;114(24):412–8.
2. Ovesen PG, Fuglsang J, Andersen MB, Wolff C, Petersen OB. David McIntyre H. Temporal Trends in Gestational Diabetes Prevalence, Treatment, and Outcomes at Aarhus University Hospital, Skejby, between 2004 and 2016. J Diabetes Res. 2018;2018:597059.
3. International Diabetes Federation. IDF Diabetes Atlas. 8th ed. Brussels: International Diabetes Federation; 2017.
4. Lee KW, Ching SM, Ramachandran V, Yee A, Hoo FK, Chia YC, et al. Prevalence and risk factors of gestational diabetes mellitus in Asia: a systematic review and meta-analysis. BMC Pregnancy Childbirth. 2018;18(1):494.
5. Jafari-Shobeiri M, Ghojazadeh M, Azami-Aghdash S, Naghavi-Beztad M, Piri R, Pourall-Akbar Y, et al. Prevalence and Risk Factors of Gestational Diabetes in Iran: A Systematic Review and Meta-Analysis. Iran J Public Health. 2015;44(8):1036–44.
6. Keshavarz M, Cheung NW, Babaei GR, Moghadam HK, Ajami ME, Shariati M. Gestational diabetes in Iran: incidence, risk factors and pregnancy outcomes. Diabetes Res Clin Pract. 2005;69(3):279–86.
7. Larrabure-Torreavl GT, Martinez S, Lune-Fernandez MA, Sanchez SE, Mascaro PA, Ingar H, et al. Prevalence and risk factors of gestational diabetes mellitus: findings from a universal screening feasibility program in Lima, Peru. BMC Pregnancy Childbirth. 2018;18(1):303.
8. Delitto CL, Kim SY, Sharma AJ. Prevalence estimates of gestational diabetes mellitus in the United States, Pregnancy Risk Assessment Monitoring System (PRAMS), 2007–2010. Prev Chronic Dis. 2014;11:E104.
9. Mu M, Xu LF, Hu D, Wu J, Bai MJ. Dietary Patterns and Overweight/Obesity: A Review Article. Iran J Public Health. 2017;46(7):869–76.
10. Jannasch F, Kröger J, Schulze MB. Dietary Patterns and Type 2 Diabetes: A Systematic Literature Review and Meta-Analysis of Prospective Studies. J Nutr. 2017;147(6):1174–82.

11. Bowers R, Tobias DK, Yeung E, Hu FB, Zhang C. A prospective study of prepregnancy dietary fat intake and risk of gestational diabetes. Am J Clin Nutr. 2012;95(2):446–53.

12. Liang Y, Gong Y, Zhang X, Yang D, Zhao Q, Quan L, et al. Dietary Protein Intake, Meat Consumption, and Dairy Consumption in the Year Preceding Pregnancy and During Pregnancy and Their Associations With the Risk of Gestational Diabetes Mellitus: A Prospective Cohort Study in Southwest China. Front Endocrinol (Lausanne). 2018;9:596.

13. Hu FB. Dietary pattern analysis: a new direction in nutritional epidemiology. Curr Opin Lipidol. 2002:13:3–9.

14. Cespedes EM, Hu FB. Dietary patterns from nutritional epidemiologic analysis to national guidelines. Am J Clin Nutr. 2015;101(5):899–900.

15. Hassani Zadeh S, Boltetta P, HosseinZadeh M. Dietary patterns and risk of gestational diabetes mellitus: A systematic review and meta-analysis of cohort studies. Clin Nutr ESPEN. 2020;36:1–9.

16. Raghavan R, Dreibelbis C, Kingshipp BL, et al. Dietary patterns before and during pregnancy and maternal outcomes: a systematic review. Am J Clin Nutr. 2019:109(Suppl_7):7055–7285.

17. Mijatovic-Vukas J, Cipling L, Cheng S, Sarmataks E, Louie J, Cheung NW, et al. Associations of Diet and Physicial Activity with Risk for Gestational Diabetes Mellitus: A Systematic Review and Meta-Analysis. Nutrients. 2018;10(6):698.

18. Schoenaker DA, Mishra GD, Callaway LK, Soedamah-Muthu SS. The Role of Energy, Nutrients, Foods, and Dietary Patterns in the Development of Gestational Diabetes Mellitus: A Systematic Review of Observational Studies. Diabetes Care. 2016;39(1):16–23.

19. DU HY, Jiang H, O K, Chen B, Xu LJ, Liu SP, et al. Association of Dietary Patterns during Pregnancy and Gestational Diabetes Mellitus: A Prospective Cohort Study in Northern China. Biomed Environ Sci. 2017;30(12):887–97.

20. Mak JKL, Pham NM, Lee AH, Tang L, Pun XF, Binns CW, et al. Dietary patterns during pregnancy and risk of gestational diabetes: a prospective cohort study in Western China. Nutr J. 2018;17(1):107.

21. Sedaghat F, Akhoondan M, Ehteshami M, Aghamohammadi V, Ghanem N, Mirmiran P, et al. Maternal Dietary Patterns and Gestational Diabetes Risk: A Case-Control Study. J Diabetes Res. 2017;2017:5173926.

22. American Diabetes Association. Diabetes management guidelines. Diabetes Care. 2015;38(Suppl 1):1–93.

23. Fleiss JL, Levin B, Paik MC. Statistical Methods for Rates and Proportions. 3rd Edition, Hoboken: John Wiley & Sons; 2003.

24. Dean AG, Sullivan KM, Soe MM. OpenEpi. Open Source Epidemiologic Statistics for Public Health, Version. www.OpenEpi.com, [Accessed 15 Jan 2018].

25. Shakeri M, Jafarirad S, Amarni R, Cheraghian B, Najaﬁ M. A longitudinal study on the relationship between mother’s personality trait and eating behaviors, food intake, maternal weight gain during pregnancy and neonatal birth weight. Nutr J. 2020;19(1):67.

26. Craig CL, Marshall AL, Sjöström M, Bauman AE, Booth ML, Ainsworth BE, et al. International physical activity questionnaire: 12-country reliability and validity. Med Sci Sports Exerc. 2003;35(8):1361–8.

27. Mirmiran P, Esfahani FH, Mehrabi Y, Hedayati M, Azizi F. Reliability and relative validity of an FFQ for nutrients in the Tehran lipid and glucose study. Public Health Nutr. 2010;13(5):564–62.

28. Willett WC, Howe GR, Kushi LH. Adjustment for total energy intake in quantitative data by quantiles in epidemiologic studies: classification according to cases, noncases, or all subjects? Epidemiology. 1991;2(2):137–40.

29. Santangelo C, Zicari A, Mandolesi F, Sciacchitano B, Mari E, Morano S, et al. Could gestational diabetes mellitus be managed through dietary bioactive compounds? Current knowledge and future perspectives. Br J Nutr. 2016;115(7):1299–44.

30. Jolliffe IT, Cadima J. Principal component analysis: a review and recent developments. Philos Trans A Math Phys Eng Sci. 2016;374(2065):20150202.

31. Pallant J. SPSS survival manual: McGraw-Hill education (UK); 2013.

32. Hsieh CC, Maissenneuve P, Boyle P, Macfarlane GJ, Roberston C. Analysis of quantitative data by quantiles in epidemiologic studies: classification according to cases, noncases, or all subjects? Epidemiology. 1991;2(2):137–40.

33. Aune D, Norat T, Romundstad P, Vatten LJ. Dairy products and the risk of type 2 diabetes: a systematic review and dose-response meta-analysis of cohort studies. Am J Clin Nutr. 2013;98(4):1066–83.

34. Asemi Z, Samimi M, Tabassi Z, Sabithi SS, Esmaillzadeh A. A randomized controlled clinical trial investigating the effect of DASH diet on insulin resistance, inflammation, and oxidative stress in gestational diabetes. Nutrition. 2013;29(4):619–24.

35. Fumeron F, Lamri A, Abi Khalil C, Jaziri R, Porchay-Baldelloul I, Lantieri O, et al. Dairy consumption and the incidence of hyperglycemia and the metabolic syndrome: results from a french prospective study. Data from the Epidemiologic Study on the Insulin Resistance Syndrome (DESIR). Diabetes Care. 2011;34(6):813–17.

36. Osorio-Yáñez C, Qiu C, Gayle B, Esguibahrie DA, Williams MA. Risk of gestational diabetes mellitus in relation to maternal dietary calcium intake. Public Health Nutr. 2017;20(6):1082–9.

37. Adams RL, Broughton KS. Insulinotropic Effects of Whey: Mechanisms of Action, Recent Clinical Trials, and Clinical Applications. Ann Nutr Metab. 2016;69(1):56–63.

38. Asadi M, Shahrezaei M, Nadjarzadeh A, Hashemi Yusufabadi H, Mansoori A. The relationship between prepregnancy dietary patterns adherence and risk of gestational diabetes mellitus in Iran: A case–control study. Nutrition. 2019;76(6):597–603.

39. Jafarirad S, Mousavi Borjazi A, Fathi M, Hormoznejad R. A Study on Anthropometric Measurements, Blood Pressure, Blood Sugar and Food Intakes Among Different Social Status and Ethnicities. Jundishapur J Chronic Dis Care. 2017;6(1):e38916.

40. Imamura F, Micha R, Khatibzadeh S, Fahimi S, Shi P, Powles J, et al. Dietary quality among men and women in 187 countries in 1990 and 2010: a systematic assessment. Lancet Glob Health. 2015;3:e132–42.

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