Association between severity of COVID-19 symptoms and habitual food intake in adult outpatients

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ABSTRACT

Objective To evaluate the association between habitual frequency of food intake of certain food groups during the COVID-19 pandemic and manifestations of COVID-19 symptoms in adult outpatients with suspected SARS-CoV-2 infection.

Design We included 236 patients who attended an outpatient clinic for suspected COVID-19 evaluation. Severity of symptoms, habitual food intake frequency, demographics and Bristol chart scores were obtained before diagnostic confirmation with real-time reverse transcriptase PCR using nasopharyngeal swab.

Results The results of the COVID-19 diagnostic tests were positive for 103 patients (44%) and negative for 133 patients (56%). In the SARS-CoV-2-positive group, symptom severity scores had significant negative correlations with habitual intake frequency of specific food groups. Multivariate binary logistic regression analysis adjusted for age, sex and occupation confirmed that SARS-CoV-2-positive patients showed a significant negative association between having higher symptom severity and the habitual intake frequency of ‘legumes’ and ‘grains, bread and cereals’ food groups.

Conclusions Increase in habitual frequency of intake of ‘legumes’, and ‘grains, bread and cereals’ food groups decreased overall symptom severity in patients with COVID-19. This study provides a framework for designing a protective diet during the COVID-19 pandemic and also establishes a hypothesis of using a diet-based intervention in the management of SARS-CoV-2 infection, which may be explored in future studies.

INTRODUCTION

Diet is significant for health and can be either a protective or a risk factor for the development of chronic diseases such as type 2 diabetes mellitus,1 hypertension,2 other endocrinological diseases3 and renal diseases.4 Diet also plays a major role in the pathogenesis of infectious respiratory diseases, such as tuberculosis5 and viral infections (ie, rhinovirus, influenza and adenovirus).6 A complex and multilayer interaction between habitual and temporal food intake, collective food consumption pattern, and health status shapes immunological and physiological status,7–9 especially in response to viral infections. As such, links can also be potentially found in infections caused by coronavirus (COVID-19). Both energy restrictions and high caloric diets can disrupt the body immune response against viral respiratory infections.10 There are also recent reports that suggest specific food groups can modulate the severity of SARS-CoV-2 infection.11 Additionally, the COVID-19 pandemic has brought substantial changes to people’s dietary habits and community food quality and availability.12 These changes are different for every region and also for individuals,13,14 which in turn can influence the immune system of the population in response to COVID-19 infection.

Nutritional status is directly associated with inflammation and consequently with immune response.10 The symptoms of the COVID-19 syndrome are variable among individuals, and the reasons that such differences are seen remain confounded15 but could potentially include diet. Besides respiratory symptoms, gastrointestinal manifestations such as nausea, vomiting, abdominal pain and diarrhoea are also relatively common in SARS-CoV-2 infections.16 However, the relations between this set of symptoms and the severity of clinical features are still unclear.16 It has
been proposed that SARS-CoV-2 infection may cause microbiome dysbiosis, which influences gastrointestinal physiology and immunological balances.15

Recent studies have begun to unravel the links between dietary microbiota modulation and host immunity. Western-style diets profoundly affect the gut microbiome configuration and impact host immunological balances.

Altogether, the type of diet highly influences the composition of the microbiota, which in turn modifies the host health.18 This intricate interaction is likely to be disrupted by SARS-CoV-2 infections and may alter the clinical course of the syndrome beyond gastrointestinal symptoms.19 Kim20 argued that unhealthy gut microorganisms might be the underlying reason for the predisposition of normal individuals to severe COVID-19. Changes in diet, especially the amount of indigestible food components such as dietary fibres, change the gut microbiota composition. David et al21 found that even short-term consumption of meals composed entirely of animal or plant products alters the microbiota structure and overwhelms individual differences in microbial gene expression.

A balanced nutrition, favouring the immune response, is essential to prevent and manage viral infections.10 Obesity, in which diet plays a major role, is associated with higher mortality rate in patients with COVID-19.22 The high prevalence of these risk factors was suggested to be driven by increased consumption of typical contemporary diets with high amounts of saturated fats, sugar and refined carbohydrates, and low levels of fibre, unsaturated fats and antioxidants.23 Detrimental changes in temporary diets with high amounts of saturated fats, sugar and refined carbohydrates, and low levels of fibre, unsaturated fats and antioxidants.23 Detrimental changes in diet can cause obesity and concurrent alterations in the human gut microbiota composition.24 In Mexico, where the study presented in this paper was conducted, 72.5% of adults are either obese and overweight, and such a high prevalence has been characterised by an excessive intake of discretionary foods.25 Moreover, during the lockdown period around the globe, meal patterns became unhealthier (ie, less fresh food, excessive eating, snacks between meals and deviation from the normal number of meals).26 A few reports have proposed that diet might explain some of the differences in mortality rates among countries.27 Although there are differences in dietary patterns, unbalanced diets are a health threat worldwide and affect death rates and quality of life.28 Such information may point to an association between food intake and clinical features of COVID-19.

The purpose of this work is to evaluate if there is a relationship between habitual intake of certain food groups and manifestations of COVID-19 symptoms in adult outpatients with suspected SARS-CoV-2 infection.

METHODS
Study protocol and participants
This cross-sectional study included 236 patients who attended an outpatient clinic for COVID-19-related evaluation in the Alvaro Obregon borough of Mexico City. The study was conducted in a period between October 2020 and January 2021. Informed consent was obtained from all participants.

All study participants were suspected cases, that is, patients who attended the clinic for either medical attention for COVID-19-related symptoms (fever, cough, odynophagia, dyspnoea, irritability, diarrhoea, chest pain, shivers, headache, myalgia, arthralgia, malaise, rhinorrhoea, polypnoea, vomiting, abdominal pain, conjunctivitis, cyanosis, anosmia and dysgeusia) or a clinical evaluation due to contact with SARS-CoV-2-positive individuals. All clinical variables were obtained during the medical interviews of the subjects with suspected cases of the viral respiratory disease. Demographic and clinical data were noted on the medical records. Patients were eligible for participation in the study if they were 18 years of age or older and capable of filling in the questionnaires. Exclusion criteria were major psychiatric or medical problems (reported on the health history), pregnancy or breast feeding. Patients were asked if they wanted to participate in the study; if the answer was positive the questionnaire was administered. Severity of symptoms and habitual food intake frequency were assessed through a questionnaire given after the clinical evaluation and before the COVID-19 diagnostic test (as detailed in the following section).

Sample size
We selected a Pearson correlation of 0.4 as the minimum to be considered relevant between food intake frequency of any food group and the overall symptom severity score. Considering a unilateral α error of 0.05 and a statistical power of 0.9, the minimum number of participants is 62. The national ratio of positive SARS-CoV-2 tests was 1:2 at the time of this study; thus, at least 186 patients were planned to be recruited during the sampling period.

Diagnostic test for SARS-CoV-2 infection
The diagnostic test for SARS-CoV-2 infection was performed using real-time PCR technique.29 Collection and analysis of the samples for diagnosis of COVID-19 were carried out by following the Standardized Guideline for Epidemiological and Laboratory Surveillance of Viral Respiratory Disease issued by the Ministry of Health, Mexico.30 A confirmed case (positive test) was a person who had a laboratory-confirmed diagnosis of COVID-19 according to the National Network of Public Health Laboratories recognised by the Institute for Epidemiological Diagnosis and Reference. Clinical evaluation and sampling procedures were carried out following the preventive measures for managing patients with viral respiratory disease for health personnel.30 Sampling was of a pharyngeal or nasopharyngeal exudate type according to the procedure described in the ‘Technical details of diagnostic tests for SARS-CoV-2 infection’ section of the online supplemental material.30 Patients’ clinical information was entered into the epidemiological surveillance system with an identification number for follow-up of case and notification of test results through the health centre.
where the sample was taken, and the case was evaluated.\textsuperscript{30} Test results were communicated to each patient by phone within 1–3 days after performing the test.

Assessment of habitual frequency of food intake
A habitual Food Frequency Questionnaire (FFQ) was designed in a web-based online form to collect information on the dietary intake of adults during the previous 3 months. It was a qualitative FFQ. The first section included 96 food items divided into 10 food groups, representing Mexico’s most commonly consumed foods. The FFQ was adapted from the Mexican FFQ ‘Frecuencia de consumo de alimentos de adolescentes y adultos 12 años o más’;\textsuperscript{31} this FFQ has been validated in Mexico by the National Institute of Public Health for collecting data during the National Health and Nutrition Survey 2018–2019. A complete list of the food items is found in online supplemental table S1.

The FFQ food groups were divided into the following categories: ‘milk and milk products’, ‘fruit’, ‘vegetables’, ‘animal products’, ‘legumes’, ‘grains, bread and cereals’, ‘corn-based products’, ‘beverages’, ‘fats and oil’, and ‘hot peppers and sauce’. Participants were asked to report the usual frequency of food item consumption in the following specified categories: ‘never’, ‘once a month’, ‘two or three times a month’, ‘once a week’, ‘two to three times a week’, ‘four to six times a week’, ‘one time per day’, and ‘two times or more per day’, which were scored from 0 to 7 in the same order.

Assessment of symptom severity
We developed a symptom severity scale for self-completion that assessed the severity of symptoms as perceived by the outpatients.\textsuperscript{32,33} Symptom severity can serve as a complementary measure of respiratory disease impact for individuals seeking care in outpatient settings.\textsuperscript{34} Symptom severity was assessed using a subjective self-rating scale of 11 symptoms: ‘fever’, ‘cough’, ‘cold chill’, ‘body pain’, ‘fatigue’, ‘difficulty of breathing’, ‘headache’, ‘sore throat’, ‘loss of taste’, ‘nausea’ and ‘diarrhea’. Participants rated each of their symptoms on a Likert scale from 1 to 5, where 1 was ‘very bad’, 2 ‘bad’, 3 ‘regular’, 4 ‘good’ and 5 ‘very good’ and scored from 0 to 5 points in the same order. Additionally, patients were asked to match the closest appearance of their stool using the Bristol chart.\textsuperscript{35}

Statistical analysis
Qualitative variables were described as absolute frequency and percentage and were compared between groups by $\chi^2$ or exact Fisher’s test. In all discrete and continuous quantitative variables, normal distribution was tested with Kolmogorov-Smirnov test. Variables with normal distribution were described as mean±SD and were compared between groups by Student’s t-test for independent groups (with variances equal or unequal according to an $F$ test). Otherwise, the variables were described as median (25th–75th percentile) and were compared between groups by Mann-Whitney U test. Spearman’s correlation analyses were performed between the total number of symptoms and the total food intake groups. Multivariate binary logistic regression analyses were performed to assess the association between having higher severity of symptoms (as a dependent variable) and frequency of food intake of each food group, age, sex and occupation (as independent variables). The results are shown as OR with 95% CI.

RESULTS
The diagnostic test was positive for SARS-CoV-2 in 103 patients (44%) and negative in 133 (56%). There were no significant differences between the groups in terms of sociodemographic characteristics (age, sex, occupation and body mass index (BMI)) as well as comorbidities (smoking, hypertension or diabetes mellitus), as shown in table 1.

The results of the analysis of symptom severity and food intake frequency are shown in table 2. Compared with patients with a negative test result, those with a positive test result had a significantly higher sum of symptoms’ severity. The median for habitual frequency of food intake was similar for patients with positive and negative test results in all food groups.

Given that symptom severity score was higher in outpatients with positive test results for COVID-19 compared with those with negative test results, we stratified the symptom severity score according to the median score of 18 points (table 2) and compared the sociodemographic characteristics and comorbidities between the two groups by test results stratified by symptom severity score (table 3). All characteristics were similar between the subgroups, except for an older age in the subgroup with negative test and symptom severity score ≤18 (compared with those with positive test and symptom severity score >18, indicated by * in table 3). Additionally, occupation was different between the subgroups with negative test results and symptom severity score >18 and those with positive test results and symptom severity score >18 (indicated by † in table 3).

As the first approach to investigate the relationships between symptoms’ severity and habitual food frequency intake, we performed bivariate correlation analyses. Table 4 shows the Spearman’s correlation analysis between symptom severity sum scores and food intake frequency. Online supplemental figure S1 shows the scatter-dot plots of the symptom severity sum score as a function of frequency of food intake of several food groups. In the group with a positive test result, symptom severity sum scores had significant negative correlations with frequency of intake of (1) ‘milk and milk products’, (2) ‘legumes’, (3) ‘grains, bread and cereals’, and (4) ‘fats
and oil’. In contrast, the group with a negative test result had significant positive correlations between symptom severity sum score and frequency of intake of (1) ‘beverages’ and (2) ‘hot peppers and sauce’.

Considering that age could be a confounding variable, we also explored the bivariate correlations between food frequency intake and age (table 4). The group with positive tests showed significant negative correlations between age and intake of (1) ‘milk and milk products’, (2) ‘fruit’, and (3) ‘beverages’. The group with negative tests showed significant correlations (positive or negative) between age and frequency of intake of most food groups: (1) ‘vegetables’, (2) ‘animal products’, (3) ‘legumes’, (4) ‘grains, bread and cereals’, (5) ‘beverages’, and (6) ‘hot peppers and sauce’.

The association between the habitual food intake and higher severity of symptoms (ie, symptom severity score >18 points) was estimated by a binary logistic regression for each food group (as independent variable), with univariate models (ie, basic models) in both groups (positive test or negative test group) (table 5). Then, age, sex and occupation were introduced in each basic model to constitute a multivariate binary regression analysis that measures the association between food intake of each group and having a high severity of symptoms, with adjustment for the confounding variables.

### Table 1 Sociodemographic characteristics and comorbidities of the study population

| Variable                        | All (N=236) | SARS-CoV-2 test result |
|---------------------------------|-------------|------------------------|
|                                 |             | Positive (n=103)       | Negative (n=133) | P value |
| **Age (years)**                 | 33 (26–45)  | 31 (26–39)             | 35 (28–46)       | 0.102   |
| Sex                             |             |                        |                  |         |
| Female                          | 153 (65)    | 70 (68)                | 83 (62)          |         |
| Male                            | 83 (35)     | 33 (32)                | 50 (38)          |         |
| **Occupation**                  |             |                        |                  |         |
| Employee                        | 93 (40)     | 35 (34)                | 58 (44)          |         |
| Housewife                       | 45 (19)     | 27 (26)                | 18 (14)          |         |
| Student                         | 26 (11)     | 11 (11)                | 16 (11)          |         |
| Other                           | 72 (31)     | 30 (29)                | 42 (32)          |         |
| **BMI (kg/m²)**                 | 27.0±5.1    | 27.1±5.7               | 27.0±4.5         | 0.879   |
| Overweight or obese             | 69 (67)     | 69 (67)                | 92 (69)          | 0.721   |
| Smoker                          | 39 (19)     | 17 (21)                | 22 (18)          | 0.526   |
| Hypertension                    | 15 (7)      | 5 (6)                  | 10 (8)           | 0.633   |
| Diabetes mellitus               | 14 (7)      | 7 (9)                  | 7 (6)            | 0.388   |

Data are shown as absolute value (percentage), mean±SD or median (25th–75th percentile). BMI, body mass index.

### Table 2 Symptom severity sum score and food intake frequency of the study population

| Variable                        | All (N=236) | SARS-CoV-2 test result |
|---------------------------------|-------------|------------------------|
|                                 |             | Positive (n=103)       | Negative (n=133) | P value |
| **Symptom severity sum score**  | 18 (12–35)  | 32 (21–48)             | 13 (11–19)       | <0.001  |
| **Food intake frequency**       |             |                        |                  |         |
| Milk and milk products          | 16±8        | 15±8                   | 16±8             | 0.414   |
| Fruit                           | 28±13       | 28±14                  | 27±12            | 0.435   |
| Vegetables                      | 44±17       | 45±18                  | 43±15            | 0.376   |
| Animal products                 | 27±8        | 28±9                   | 26±8             | 0.192   |
| Legumes                         | 7±4         | 8±4                    | 7±4              | 0.244   |
| Grains, bread and cereals       | 26±10       | 25±10                  | 27±10            | 0.249   |
| Corn-based products             | 16±7        | 16±6                   | 16±7             | 0.847   |
| Beverages                       | 21±11       | 23±10                  | 20±11            | 0.105   |
| Fats and oil                    | 14±7        | 13±7                   | 14±7             | 0.649   |
| Hot peppers and sauce           | 13±7        | 14±7                   | 13±7             | 0.363   |

Data are shown as mean±SD or median (25th–75th percentile).
In the positive test result group, the basic regression models showed that the risk of high symptom severity decreased with intake of ‘legumes’, and ‘grains, bread and cereals’ (OR below 1.0) as the most significant contributors. In the group with negative test results, the basic regression model showed that the risk of high symptom severity increased with the habitual intake of ‘fats and oil’, as well as with intake of ‘hot peppers and sauce’. As the underlying health disorders for the group with negative tests are unknown, we cannot further discuss them.

Figure 1 shows the results on appetite score, food taste score and Bristol chart score. Patients with a positive SARS-CoV-2 test had lower appetite and food taste scores than patients with a negative SARS-CoV-2 test. The Bristol chart score was similar between the groups.

### Table 3: Sociodemographic characteristics and comorbidities of the study population stratified by severity of symptoms

| Variable          | All (N=236) | SARS-CoV-2 test result | |
|-------------------|-------------|-------------------------|---|
|                   | Symptom severity score ≤18 | Symptom severity score >18 | |
|                   | | | SARS-CoV-2 test result | | | |
|                   | | | Positive (n=103) | | | Negative (n=133) |
|                   | Age (years) | | 34 (26–46) | 32 (27–44) | 28 (22–38) | 32 (26–44) | 36 (28–46)* | 32 (27–45) |
|                   | Sex | | | | | | |
| Female | 72 (61) | 81 (69) | 14 (70) | 56 (68) | 58 (59) | 25 (74) | | |
| Male | 47 (39) | 36 (31) | 6 (30) | 27 (33) | 41 (41) | 9 (26) | | |
|                   | Occupation | | | | | | |
| Employee | 56 (47) | 37 (32) | 8 (40) | 27 (33) | 48 (49) | 10 (29)† | | |
| Housewife | 19 (16) | 26 (22) | 4 (20) | 23 (28) | 15 (15) | 3 (9)† | | |
| Student | 13 (11) | 13 (11) | 5 (25) | 6 (7) | 8 (8) | 7 (21)† | | |
| Other | 31 (26) | 41 (35) | 3 (15) | 27 (33) | 28 (8) | 14 (41)† | | |
|                   | BMI (kg/m²) | | 26.8±4.8 | 27.2±5.3 | 27.0±7.6 | 27.1±5.2 | 26.8±4.0 | 27.4±5.8 |
| Overweight or obese | 79 (66) | 82 (70) | 11 (55) | 58 (70) | 68 (69) | 24 (70) | | |
| Smoker | 17 (15) | 22 (23) | 4 (21) | 13 (21) | 13 (14) | 9 (28) | | |
| Hypertension | 6 (5) | 9 (10) | 0 (0) | 5 (8) | 6 (6.4) | 4 (13) | | |
| Diabetes mellitus | 5 (4) | 9 (10) | 0 (0) | 7 (11) | 5 (5) | 2 (6) | | |

Data are shown as absolute value (percentage), mean±SD or median (25th–75th percentile).

*P<0.05 (compared with group with positive test and severity score ≤18).
†P<0.05 (compared with group with positive test and severity score >18).
BMI, body mass index.

### Table 4: Spearman’s correlation analyses between symptom severity score and age (years) versus habitual frequency of food intake

| Food group                      | SARS-CoV-2 test result | | |
|---------------------------------|-------------------------|---|---|
|                                 | Positive (n=103) | Age | Negative (n=133) | Age |
|                                 | Severity of symptoms | | Severity of symptoms | |
|                                 | P value | r | P value | r | P value | r |
| Milk and milk products | −0.213 | 0.031 | −0.260 | 0.008 | −0.050 | 0.567 | −0.166 | 0.056 |
| Fruit | −0.086 | 0.388 | −0.203 | 0.040 | −0.072 | 0.412 | 0.116 | 0.183 |
| Vegetables | −0.058 | 0.562 | 0.032 | 0.753 | −0.032 | 0.718 | 0.252 | 0.004 |
| Animal products | −0.167 | 0.092 | −0.185 | 0.062 | 0.013 | 0.883 | −0.234 | 0.007 |
| Legumes | −0.296 | 0.002 | −0.083 | 0.407 | 0.040 | 0.650 | 0.277 | 0.001 |
| Grains, bread and cereals | −0.246 | 0.025 | 0.030 | 0.789 | 0.087 | 0.320 | −0.143 | 0.100 |
| Corn-based products | −0.165 | 0.095 | −0.048 | 0.629 | 0.100 | 0.254 | −0.183 | 0.135 |
| Beverages | −0.038 | 0.703 | −0.258 | 0.009 | 0.171 | 0.049 | −0.250 | 0.004 |
| Fats and oil | −0.197 | 0.046 | −0.009 | 0.927 | 0.116 | 0.183 | −0.142 | 0.104 |
| Hot peppers and sauce | −0.078 | 0.433 | −0.085 | 0.396 | 0.225 | 0.009 | −0.234 | 0.007 |
BMI had no correlations with symptom severity sum score in patients with a positive test ($r=0.202$, $p=0.069$) and a negative test ($r=-0.103$, $p=0.877$). However, BMI had a significant correlation with age in patients with a positive test but no correlation in patients with a negative test, as shown in figure 2.

**DISCUSSION**

In general, previous studies have shown that diets high in fibre, carbohydrates and nutrients can decrease the risk of respiratory infections, such as the common cold and pneumonia, and also shorten their duration.\(^{36}\) It is generally assumed that such diets can support the immune system by stimulating the production of antibodies and lymphocytes and by reduction of oxidative stress. Diet also changes the type and frequency of the commensal microbiota. There have been suggestions regarding the impact of diet on SARS-CoV-2 infections as a result of gut microbiota modulation.\(^{10,37-39}\) Yeoh *et al.*\(^{40}\) have shown that the gut microbiome composition could be significantly altered in patients with COVID-19 compared with individuals without COVID-19, irrespective of whether patients had received medication ($p<0.01$). They have also demonstrated the associations between gut microbiota composition, cytokine levels and inflammatory markers in patients with COVID-19, suggesting that the gut microbiome make-up may be associated with COVID-19 severity.\(^{39}\)

Our study showed that there are significant correlations between severity of symptoms and habitual food intake among outpatients with SARS-CoV-2 infection. Altogether, the relationship between long-term dietary components before infection and SARS-CoV-2 symptom severity appeared to be multifactorial. While symptom severity sum score had significant bivariate negative correlations with frequency of intake of ‘milk and milk products’, ‘legumes’, ‘grains’, ‘bread and cereals’, and ‘fat and oils’, the strongest negative correlation was obtained for legumes ($r=-0.296$, $p=0.002$), table 4. The subsequent multivariate regression analyses (table 5) confirmed that, after being adjusted by confounding variables, there was a significant association between having a high symptom severity score and the habitual frequency intake of ‘legumes’ and ‘grains bread and cereals’.

These results suggest that long-term habitual intake of ‘legumes’, and ‘grains, bread and cereals’ food groups can be beneficial in protecting adults against the symptoms of SARS-CoV-2 infection. Interestingly, reliance on other food groups such as fruit and vegetables, which are frequently suggested as beneficial during COVID-19, did not appear to have a statistically significant impact on the severity of symptoms in the positive population.

Our observations are, in part, in agreement with previous studies on the role of diet in enhancing systemic immunity against SARS-CoV-2 infections. In a recent study, Kim *et al.*\(^{41}\) have shown that the severity of symptoms in patients with COVID-19 who followed a plant-based diet has been lower than those who reported following low-carbohydrate, high-protein diets. Altogether, people with a healthier nutritional status presented lower susceptibility to severe COVID-19.\(^{40}\)

Natural milk and milk products are high in beneficial fats, and milk-based products such as natural yoghurt are made of beneficial bacteria. A new study has shown that natural milk contains biocomponents that are highly protective against COVID-19 infections.\(^{41}\) Dairy is a good source of branched-chain fatty acids (BCFAs). Animal-based studies have shown that BCFAs alter the microbiota composition in micro-organisms that use these fatty acids in cellular membranes.\(^{42}\) Because they are normal

| Food group                        | SARS-CoV-2 test result |          |          |          |          |
|-----------------------------------|------------------------|----------|----------|----------|----------|
|                                   | Positive (n=103)       |          |          |          |          |
|                                   | Basic model            | Multivariate model† | Basic model | Multivariate model† |
| Milk and milk products            | 0.95 (0.89 to 1.00)    | 0.95 (0.89 to 1.01) | 0.99 (0.94 to 1.05) | 1.00 (0.94 to 1.05) |
| Fruit                             | 0.99 (0.96 to 1.03)    | 1.00 (0.96 to 1.03) | 0.97 (0.95 to 1.02) | 0.97 (0.94 to 1.01) |
| Vegetables                        | 0.99 (0.97 to 1.02)    | 1.00 (0.97 to 1.03) | 0.99 (0.97 to 1.02) | 1.00 (0.96 to 1.02) |
| Animal products                   | 0.97 (0.92 to 1.01)    | 0.97 (0.92 to 1.03) | 1.02 (0.97 to 1.08) | 1.02 (0.96 to 1.08) |
| Legumes                           | 0.87 (0.76 to 0.99)*   | 0.86 (0.74 to 0.99)* | 1.00 (0.90 to 1.12) | 1.04 (0.92 to 1.18) |
| Grains, bread and cereals         | 0.92 (0.86 to 0.98)**  | 0.91 (0.84 to 0.97)** | 1.03 (0.99 to 1.07) | 1.03 (0.99 to 1.07) |
| Corn-based products               | 0.97 (0.90 to 1.06)    | 0.98 (0.90 to 1.07) | 1.04 (0.99 to 1.10) | 1.04 (0.98 to 1.10) |
| Beverages                         | 0.99 (0.94 to 1.04)    | 0.99 (0.94 to 1.05) | 1.03 (0.99 to 1.07) | 1.03 (0.99 to 1.07) |
| Fats and oil                      | 0.95 (0.89 to 1.03)    | 0.95 (0.88 to 1.02) | 1.06 (1.00 to 1.12)* | 1.06 (0.99 to 1.12) |
| Hot peppers and sauce             | 0.96 (0.90 to 1.03)    | 0.96 (0.90 to 1.03) | 1.06 (1.00 to 1.11)* | 1.07 (1.01 to 1.14)* |

Results are shown as OR (95% CI).

*P< 0.05, **P<0.01.
†Multivariate models were adjusted for age, sex and occupation.
constituents of the healthy human infant gut, these fatty acids could play a role in normal colonisation. ACE2 receptor is a key target of SARS-CoV-2. Peptides found in milk pose an ACE inhibitory activity. Probiotics such as Bifidobacteria and Lactobacillus which are found in dairy products are beneficial in building a resilient and diversified gut microbiota. Additionally, probiotics such as Lactobacillus acidophilus and L. casei which are found in natural yoghurt are suggested to boost systemic immunity against SARS-CoV-2 infection. In comparison, in our work, the effect of ‘milk and milk products’ and ‘fats and oil’ habitual intakes were inconclusive, since the association with symptom severity score was not significant after adjusting for the confounding variables (table 5).

A very important finding of this work was that patients with positive COVID-19 tests and mild symptoms had a high intake of legumes in particular, and grains, bread and cereals to a lesser extent. Legumes are among the most consumed food groups among the population of this study (who live in Mexico) and provide soluble fibre and amino acids. Legumes are a good source of branched-chain amino acids, which increase intestinal immunoglobulin level, improving the gut barrier. Non-digestible starch (known as resistant starch), mainly present in heat-treated grains, is considered a significant part of dietary fibres in many countries including Mexico. It is known that resistant starch increases short-chain fatty acid (SCFA) production in the colon, including propionate, acetate and butyrate, resulting from the fermentation of undigested soluble dietary fibres by the gut microbiota. These are important immunomodulatory metabolites, and their role has been extensively explored in the gut–lung axis. Portune et al. have reported an increase in beneficial bacteria such as Ruminococcus spp and Bifidobacterium spp due to a relative increase in dietary fibre, which is associated with enhanced systemic immunity. Plant-based fibre also has prebiotic effects, promoting the growth of bacteria that are associated with health benefits, such as Bifidobacterium spp and Lactobacillus spp. A fibre-rich diet modifies both the lung and gut microbiota composition and protects against allergic inflammation through increased circulating levels of SCFAs. Similarly, an increase in the population of Bifidobacterium and Lactobacillus was also seen among people on a bread-based diet. In contrast, in populations with reduced polysaccharide intake, potentially opportunistic pathogens are increased, such as Escherichia coli and total Enterobacteriaceae. A recent study by Merino et al. found that a dietary pattern characterised by healthy plant foods was associated with lower risk and severity of COVID-19.

Yeoh et al. have shown that specific gut commensals with known immunomodulatory potential, including Faecalibacterium prausnitzii, Eubacterium rectale and Bifidobacteria, were under-represented in patients with COVID-19. The two most dominant bacterial species in the human colon are F. prausnitzii and E. rectale, contributing significantly to butyrate production. It is known that legumes, bread and cereals promote the presence of F. prausnitzii, E. rectale and other beneficial flora in the gut, thus contributing to the overall health of the host.
rectale and Bifidobacteria, which produce SCFAs. Higher bifidobacterial counts and higher total SCFAs are related to greater intake of plant-based nutrients, such as vegetable proteins and polysaccharides.40

Interestingly, we found that the habitual intake of legumes and grains, bread and cereals food groups did not show any effect in outpatients with negative COVID-19 tests. This may indicate that the effect of these types of foods is only in COVID-19-positive patients and their habitual intake may not decrease symptom severity in other patients, such as those with the common cold or influenza. In a recent study, Gu et al66 showed that the gut microbial composition of patients with COVID-19 was different from that of H1N1 patients and healthy controls. One of their findings was that the abundance of Prevotella, Eckiella, Murdochiella and Porphyromonas was higher in the H1N1 group than in patients with COVID-19. This is an observation that should be further explored in future studies.

This study was conducted in the Alvaro Obregon borough of Mexico City, Mexico. As such, habitual accessible food groups are limited to the region. There may be a limitation as food grouping includes regional foods. However, food groups definitions and categorisations may be different in other populations. This cross-sectional investigation shows a correlation between food intake patterns and symptom severity, while it does not necessarily determine causality. Some confounders have been addressed, including age, sex and occupation. Nevertheless, other variables that could influence the outcomes of this work remain to be studied, for instance SARS-CoV-2 strain type and the genetic and immunological features of this population that might impact clinical features. Furthermore, individual characteristics and community behaviours may modify the perception of severity of symptoms and hence the search for diagnostic and treatment services.

In other studies, it has been observed that the nutritional quality of food diminished during the lockdown periods,51 as well as physical activity.52 However, this phenomenon could not be detected in our population due to lack of a previous comparison. Also, it is important to mention that the study was conducted before the dominance of the SARS-CoV-2 delta strain in Mexico, which may alter the outcomes. Although this is an open-population study, the clinic where the patients were treated offers medical assistance to mainly the inhabitants of 1 of the 16 boroughs located in Mexico City. Consequently, these findings may be different in other populations. The food patterns are specific to the population investigated, so the contribution of findings from this single study is limited and replicating the study findings in other populations is important. Finally, total caloric intake was not included in the analysis as we employed a non-quantitative FFQ.

In conclusion, the potential relationship between habitual intake of certain food groups and symptom severity in suspected SARS-CoV-2 infections was explored in 236 adults outpatients, with either positive or negative PCR test result. They attended an outpatient clinic for suspected COVID-19 and then answered a habitual FFQ. Out of 236 outpatients, diagnostic tests were positive for 103 and negative for 133. SARS-CoV-2-positive outpatients showed a significant negative correlation between symptom severity and habitual frequency of intake of ‘legumes’, and ‘grains, bread and cereals’ food groups. As a result, it may be concluded that an increase in habitual frequency of intake of such food groups may decrease the severity of symptoms in those infected with SARS-CoV-2. This work can be potentially used as a guide for designing suitable diets during the COVID-19 pandemic and also for establishing a future vision that diet-based interventions may be explored for mitigating COVID-19 severity.

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