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Relationship between dietary pattern and depressive symptoms: an international multicohort study

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Abstract

Background Several previous studies have shown that dietary patterns are associated with the incidence of depressive symptoms. However, the results have been inconsistent. This study aimed to prospectively investigate the association between dietary patterns and the risk of depressive symptoms in two large cohort studies.

Methods The Tianjin Chronic Low-grade Systemic Inflammation and Health (TCLSIH) cohort study included a total of 7,094 participants living in Tianjin, China from 2013 to 2019, and the UK Biobank cohort study includes 96,810 participants who were recruited from 22 assessment centers across the UK taken between 2006 and 2010. All participants were free of a history of cardiovascular disease (CVD), cancer, and depressive symptoms at baseline. Dietary patterns at baseline were identified with factor analysis based on responses to a validated food frequency questionnaire in TCLSIH or Oxford WebQ in UK Biobank. Depressive symptoms were evaluated using the Chinese version of the Zung Self-Rating Depression Scale (SDS) in TCLSIH or hospital inpatient records in UK Biobank. Cox proportional hazards regression models were used to estimate the association between dietary patterns and depressive symptoms.

Results A total of 989, and 1,303 participants developed depressive symptoms during 17,410 and 709,931 personyears of follow-up. After adjusting for several potential confounders, the multivariable HRs (95% CIs) of the depressive symptoms were 0.71 (0.57, 0.88) for traditional Chinese dietary pattern, 1.29 (1.07, 1.55) for processed animal offal included animal food dietary pattern, and 1.22 (1.02, 1.46) for sugar rich dietary pattern in TCLSIH (all Q4 vs Q1). In the UK Biobank, the HRs (95% CIs) of depressive symptoms were 1.39 (1.16, 1.68) for processed food dietary pattern (Q4 vs Q1), 0.90 (0.77, 1.00) for healthy dietary pattern (Q3 vs Q1), and 0.89 (0.75, 1.05) for meat dietary pattern (Q4 vs Q1) in the final adjusted model.

Conclusion Dietary patterns rich in processed foods were associated with a higher risk of depressive symptoms, and following a traditional Chinese dietary pattern or healthy dietary pattern was associated with a lower risk of depressive symptoms, whereas meat dietary pattern was not associated.

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Keywords Dietary pattern, Depressive symptoms, Cohort studies, Adults

Introduction

Depression is a common but serious mental health disorder, with an estimated 3.8% of the population affected [1], which has become a major factor in the decline in quality of life and is also associated with premature death [2]. It is estimated that the current treatment can only reduce the disease burden of patients with depressive symptoms by one-third [3]. The public health burden associated with depressive symptoms may be reduced through the identification of modifiable lifestyle factors (e.g., dietary factors) for early prevention.

Previous studies have demonstrated that some single foods or nutrients may affect the incidence of depressive symptoms [4–6]. However, people do not eat isolated nutrients or food. In addition, the food they eat may consist of a variety of kinds that provide complex combinations of nutrition for the human body. Due to the interaction and synergy among nutrients, single food or nutrient studies can't well reflect the overall effect of food combinations. Therefore, the present study considered the whole diet rather than individual nutrients or foods to examine the complex association between dietary factors and the risk of depressive symptoms [7].

To date, results from several previous cross-sectional studies [8-10] and case-control studies [11, 12] in adults investigating the association between dietary patterns and depressive symptoms are mixed; however, these studies have methodological limitations, such as the cross-sectional design. In addition, few prospective studies have shown that healthy dietary pattern was inversely associated with the risk of depressive symptoms, whereas the Western dietary pattern was positively associated with the risk of depressive symptoms [13–15]. In contrast, a prospective study of US females did not find an association between the prudent or Western pattern and depressive symptoms risk [16]. Of note, these limited prospective studies were conducted in Western countries, so their results may not be generalizable to Chinese populations with different eating habits and lifestyles. Furthermore, dietary habits and incidence of depressive symptoms within distinct races were different [17, 18]. Therefore, we examined the associations of dietary patterns with the incidence of depressive symptoms in two large cohort studies: the Tianjin Chronic Low-grade Systemic Inflammation and Health (TCLSIH) and the UK Biobank.

Methods

Study population

The TCLSIH cohort study is an ongoing dynamic prospective cohort that began in 2007. The enrolled participants included men and women aged 18-90 years living in Tianjin, China, for at least 5 years. In this study, the participants were randomly recruited while having their annual health examinations at the Tianjin Medical University General Hospital-Health Management Center, the largest and most comprehensive physical examination center in Tianjin. The exclusion criteria were disability, severe mental illness, severe cognitive impairment, hearing impairment, and pregnancy. The overall response rate was 97.3%. Participants in this study covered nearly all occupations as well as retired residents. Examinations were usually performed in the same month every year, and participants were asked to complete a structured selfadministered health status questionnaire according to their actual situation while having health examinations. We can obtain information about their socioeconomic status, mental health, physical activity, food intake, in addition to other related information through questionnaires. The Institutional Review Board of Tianjin Medical University has approved this study's protocols and procedures (reference number: TMUh-MEC 201,430). Besides, all the participants have written informed consent.

The eligible participants were followed from May 2013 to December 2019. During the study period, 11,136 adults who had received at least one health examination and returned the questionnaire. We excluded those with a history of cardiovascular disease (CVD) (n=486), cancer (n=75), had missing data or extreme total energy intake [<542 (below the 2.5 percentile) or>5,721 (over the 97.5 percentiles) kcal/day for men, and <660 (below the 2.5 percentile) or>4,947 (over the 97.5 percentiles) kcal/day for women; n=913], or had depressive symptoms at baseline (n=1,646). Besides, the participants who did not finish follow-up health examinations were excluded (n=922, retention rate: 88.5%). Finally, 7,094 participants were included in the cohort analysis (Fig. 1).

The UK Biobank is a prospective population-based cohort study that recruited half a million men and women (39–72 years old) from the general population between 2006 and 2010 [19]. The UK Biobank protocol is available online (http://www.ukbiobank.ac.uk/wp-content/uploads/2011/11/UK-Biobank-Protocol.pdf). Participants had to be registered with a general practitioner and live within 25 miles of an assessment center



Fig. 1 Selection of study participants in the TCLSIH Cohort

(England, Wales, and Scotland) to take part. Participants provided extensive information through questionnaires, interviews, health records, physical measurements, and samples of blood and urine. The UK Biobank Study was approved by the North West Multi-centre Research Ethics Committee (REC reference for UK Biobank 11/ NW/0274). At the touchscreen, all participants gave informed consent using a signature-capture device.

In the UK Biobank, participants who have completed at least two web-based 24-h dietary assessments were assessed (n=126,855). We excluded those with a history of CVD (n=3,160), cancer (n=20,562), and depressive symptoms (n=820) before the first diet recall. We also excluded participants who had unreliable total energy estimated intake data [<800 or>4,200 kcal/day for men, and<600 or>3,500 kcal/ day for women; n = 5,503], a total of 96,810 participants were finally included in our study (Fig. 2).

Assessment of depressive symptoms

In the TCLSIH cohort study, we used the Chinese version of the Zung Self-Rating Depression Scale (SDS) to evaluate depressive symptoms [20]. This scale was designed by Zung in 1965. The SDS has 20 items, each item has 1 to 4 points. The total score ranges from 20 to 80, and a higher score indicates depressive symptoms were serious. We set the cutoff score to 45 in our present study, a score higher than that can be considered depressive symptoms. We conducted a screening test to calculate the sensitivity and specificity of the SDS in the study population (120 patients with clinically diagnosed depressive symptoms and 120 healthy controls). When the SDS score of



Fig. 2 Selection of study participants in the UK Biobank

45 was used as a cutoff and the Diagnostic and Statistical Manual of Mental Disorders (fourth edition) was used as the criteria for depressive symptoms, the sensitivity and specificity were 83.6% and 96.4%, respectively [12].

In the UK biobank, diagnoses of depressive symptoms were recorded using the International Classification of Diseases (version 10; code ICD-10) coding system. Participants with depressive symptoms were identified as having a primary/secondary diagnosis (hospital records) using ICD-10 codes.

Dietary assessment

In the TCLSIH cohort study, we used the food frequency questionnaire (FFQ) which included 100 food items with specified serving sizes to assess the dietary intake since May 2013. Each food in this FFQ has seven frequencies (ranging from "almost never" to "two or more times/day") and eight frequencies for beverages consumed (ranging from "almost never" to "four or more times/day"). All participants were asked how often they had consumed each food by choosing from predefined frequency categories on average last month. The reproducibility and validity of the questionnaire were assessed in a random sample of participants from our cohort using data from repeated measurements of the FFQ approximately 3 months apart and 4-day weighed diet records (WDRs). The participants for the validation were randomly selected from different subgroups (age: 20-30, 30-40, 40-50, 50-60, 60-70, and >70 years) of the TCLSIH cohort study participants, and at least 10 men and 10 women were included in each of these subgroups. The Spearman correlation coefficients between the two FFQs were 0.68 for energy intake, and 0.62-0.79 for food items (e.g., fruits, vegetables, and beverages). The Spearman correlation coefficients between the gold standard (WDRs) and FFQs were 0.49 for total energy intake, 0.39-0.72 for energy-adjusted nutrients (e.g., vitamin C, vitamin E, polyunsaturated fatty acid, saturated fatty acids, carbohydrates). The total energy and nutrient intake (kcal/day) was calculated by using an ad hoc computer program developed to analyze the FFQ and the nutrient database derive from the Chinese Food Composition Tables [21]. We categorized the food items and beverages from the FFQ into 30 predefined food groups,

In the UK Biobank, dietary information was collected using the Oxford WebQ, a web-based self-administered 24 h dietary assessment questionnaire developed for using in large population studies. UK Biobank participants were invited up to five times to complete the Oxford WebQ dietary assessment, and we calculated mean values from the available data. The Oxford WebQ dietary assessment asks about the consumption of over 200 common food and beverage items in the previous 24 h [22]. We created new variables for the weight (gram) from the 24-h dietary assessments to estimate the daily intake of each food or drink. The top frequency category was open-ended so we coded 3 + = 3, 4 + = 4, 5 + = 5, 6 + = 6 and less than one was coded as 0.5. Weight was generated by multiplying the frequency of consumption by the serving size in grams. Finally, we classified foods and beverages reported in the Oxford WebQ into 14 main food categories (12 food and 2 beverage categories, Spreads, sauces, and cooking oil were excluded) based on a previous study [23].

Identification of dietary patterns

Due to the differences in food selection within distinct races, we categorized the food items and beverages from the TCLSIH cohort into 30 predefined food groups and 14 food groups in the UK Biobank cohort based on nutrient and culinary similarities [23]. Dietary patterns were identified using factor analysis (principal component analysis) based on the food groups in these two cohorts. The factors were rotated by an orthogonal transformation (varimax rotation function in SAS software) to achieve a simple structure, allowing greater interpretability. After combining eigenvalues criteria (>1.5), Scree plot test, greater interpretability of factors, and percentage of variance explained by the factors, factors were named descriptively based on food items with high factor loading. In the TCLSIH cohort study, three factors were named traditional Chinese dietary pattern, processed animal offal included animal food dietary pattern, and sugar rich dietary pattern. And in the UK Biobank study, three factors were named as healthy dietary pattern, processed food dietary pattern, and meat dietary pattern. Factor loadings represented the correlation coefficient between food groups and specific dietary patterns, where positive loadings represented positive correlation and negative represented inverse correlation. The dietary pattern score for each pattern was calculated by summing observed intakes of food groups weighted by their factor loadings. A higher score suggested greater adherence to a certain dietary pattern. Dietary pattern scores were standardized to have a mean of 0 and a standard deviation of 1. A higher dietary pattern score means greater conformity to this dietary pattern, contrarily unrelated.

Assessment of baseline covariates

In TCLSIH, sociodemographic characteristics (age, sex, education level, occupation, monthly household income, living alone, and the frequency of visiting friends), lifestyle, family and personal disease histories, and current medicines were obtained from a detailed and structured health-related questionnaire. All participants received a standardized physical examination, to obtain the height and weight data. Body mass index (BMI, kg/m²) was calculated using weight divided by height square. Physical activity (PA) in recent weeks was assessed by using the short form of the International Physical Activity Questionnaire [24]. Total PA levels were calculated according to the following formula: metabolic equivalent (MET) × hour/week.

All the participants were asked to fast overnight before collecting blood samples. Total cholesterol (TC), total triglycerides (TG), low-density lipoprotein cholesterol (LDL-C), and high-density lipoprotein cholesterol (HDL-C) were measured using an automatic biochemistry analyzer (Roche Cobas 8000 modular analyzer, Mannheim, Germany). Fasting blood glucose (FBG) was measured by enzymatic methods; glycated hemoglobin (HbA1c) was measured by high-performance liquid chromatographic method (HLC-723 G8; Tosoh, Tokyo, Japan); 2-h serum glucose was measured at 2 h after administration during the oral glucose tolerance test. Diabetes was defined as FBG level \geq 7.0 mmol/L, and/or 2-h serum glucose \geq 11.1 mmol/L, and/or HbA1c \geq 48 mmol/mol (6.5%), and/or taking hypoglycemic drugs [25]. Blood pressure (BP) was measured on the upper left arm with an automatic device (TM-2655P, A&D Company, Ltd., Tokyo, Japan). The participants' BP was measured twice and calculated the mean of the two data. Hypertension was defined as systolic/diastolic blood pressure \geq 140/90 mm Hg or having a history of hypertension or use of antihypertension medications [26]. Hyperlipidemia was defined as TC \geq 5.17 mmol/L, and/or TG \geq 1.7 mmol/L, and/or LDL \geq 3.37 mmol/L, and/or taking lipid-lowering drugs.

In the UK Biobank, the baseline questionnaire was used to assess several possible confounding variables including age, sex, smoking status, alcohol drinking status, whether they are living alone, and the frequency of visiting friends. We also collect information about comorbidities and family history of the disease. Townsend deprivation index was derived from the postcode of residence using aggregated data on unemployment, car and homeownership, and household overcrowding [27].

Statistical analysis

The normality of the continuous variables was examined by the Kolmogorov–Smirnov test (n > 2000). Baseline characteristics of the study participants by depressive symptoms status were presented as geometric mean and 95% confidence intervals (CIs) for continuous variables and percentages for categorical variables. The differences between participants with and without incident depressive symptoms were examined by using the *T* test for continuous variables, and the χ^2 test for proportional variables.

Person-years of follow-up for each participant in TCLSIH were calculated from the date when the participants first completed their FFQ survey and ended at the date of the first diagnosis of depressive symptoms, the end of follow-up (31 December 2019), or loss to followup, whichever was earliest. Follow-up time in the UK Biobank was calculated from the date of first completed the Oxford WebQ dietary assessment until the date of incident depressive symptoms, death, loss to follow-up, or end of study (January 31, 2018). Cox proportional hazards regression models were used to evaluate the relationship between the dietary pattern and the risk of depressive symptoms. The proportional hazards assumption was tested by Schoenfeld residuals and no violation of this assumption was found in our analyses. In further analysis, the dietary pattern score was set as independent variable in quartile categories and depressive symptoms were set as the dependent variable. Three models were fitted for the outcomes. Model 1 was adjusted for age, sex, and BMI. Model 2 was additionally adjusted for smoking status, alcohol drinking status, education levels, household income (only in the TCLSIH cohort), employment status (only in the TCLSIH cohort), Townsend deprivation index (only in the UK Biobank), total energy intake, physical activity, frequency of visiting friends, living alone based on model 1. Model 3 was additionally adjusted for family history of diseases (hypertension, CVD, hyperlipidemia (only in TCLSIH cohort), and diabetes), and individual history of the disease (hypertension, hyperlipidemia (only in TCLSIH cohort), and diabetes). We also did sensitivity analyses by including a total of 163,476 participants who completed ≥ 1 webbased 24-h dietary assessment and were without CVD, cancer, and depressive symptoms at baseline in the UK Biobank cohort.

Hazard ratios with their corresponding 95% CIs were calculated and two-sided P < 0.05 was considered as statistically significant. The linear trend test was performed by assigning medians to each quintile as a continuous variable in the model. The median value for each quartile was used to test for trends across quartiles. The likelihood ratio test was used to assess the significance of the

variables. The statistical analyses were performed by SAS version 9.3 (SAS Institute Inc., Gary, NC, USA).

Results

Tables 1 and 2 show the top 15 (TCLSIH) and all UK Biobank food items and factor loads for each dietary pattern extracted by factor analysis. In the TCLSIH cohort study, these three patterns explained 31.3% of the variance in dietary intake. The first pattern was characterized by a high intake of vegetables and grains, therefore, was identified as a traditional Chinese dietary pattern. The second pattern consisted of a high intake of organ meat, red meat, processed meat, and preserved food named processed animal offal included animal food dietary pattern. The third pattern had high factor loadings for dim sum, fruits, desserts, and sweetened beverages, and was named the sugar rich dietary pattern. In the UK Biobank study, the three dietary patterns were named healthy dietary pattern, processed food dietary pattern, and meat dietary pattern according to the factor loadings of the food group. These three patterns explained 31.9% of the variance in dietary intake.

The baseline characteristics of the study participants from the TCLSIH cohort and UK Biobank are shown in Table 3. In the TCLSIH cohort study, 7,094 participants were enrolled, and within 17,410 person-years of followup, 989 participants developed depressive symptoms. In UK Biobank, 96,810 participants were enrolled, and within 709,931 person-years of follow-up, 1303 participants developed depressive symptoms.

In brief, in the TCLSIH cohort, compared with subjects who did not develop depressive symptoms, those participants with incident depressive symptoms tended to have a lower proportion of education level, visiting friends, and household income (all P values < 0.05). In the UK Biobank, compared with subjects who did not develop depressive symptoms, those participants with incident depressive symptoms tended to have a higher BMI, Townson deprivation index, and a higher proportion of females, living alone, hypertension, diabetes, and a family history of CVD.

In Table 4, cox proportional hazards regression models show the association between dietary patterns and the risk of depressive symptoms. In the TCLSIH cohort study, after adjusting for demographic factors, lifestyle factors, disease history, and each other dietary pattern score, the final multivariate model shows that participants in the highest quartile of traditional Chinese dietary pattern score had a multivariable HR of 0.71 (95% CI: 0.57, 0.88; *P* for trend < 0.001), compared with those in the lowest quartile. For the same comparison, participants in the highest quartile of processed animal offal included animal food dietary pattern score **Traditional Chinese dietary pattern**

Food items	Factor loadings	Food items	Factor loadings
Processed animal offa dietary pattern	l included animal food	Sugar rich dietary p	attern
ed from factor analysis	s (principal component	analysis) in the TCLS	IH Cohort Study ^a

Table 1	Factor loadings for food iten	is derived from factor a	analysis (principal com	ponent analysis) in the T	CLSIH Cohort Study ^a

Food items	Factor loadings	Food items	Factor loadings	Food items	Factor loadings
Solanaceous vegetables	0.72	Organ meats	0.68	Dim sum (Cakes, cookies, and biscuits)	0.56
Dark green and leafy veg- etables	0.67	Red meats	0.55	Fruits (all kinds except for persimmon, strawberry, and kiwi fruit)	0.54
Starchy vegetables	0.65	Animal blood	0.54	Fruits (persimmon, strawberry, kiwi fruit)	0.53
Mushroom and fungi	0.61	Steamed stuffed bun, dump- ling, and wonton	0.52	Sweets and desserts	0.49
Legumes and soy products	0.58	Preserved food	0.52	Soda and sweetened bever- ages	0.43
Whole grains	0.48	Noodles	0.49	Bread	0.37
Refined grains	0.48	Sausage	0.48	Nuts	0.34
Allium vegetables	0.47	Poultry	0.46	Coffee	0.33
Chicken egg	0.46	Fish and Seafood	0.42	Salted food	0.33
Fruits (all kinds except for persimmon, strawberry, and kiwi fruit)	0.42	Salted food	0.36	Preserved food	0.28
Nuts	0.37	Soda and sweetened bever- ages	0.31	Starchy vegetables	0.28
Poultry	0.33	Alcohol consumption	0.29	Allium vegetables	0.27
Fish and Seafood	0.33	Allium vegetables	0.24	Dairy products	0.26
Red meats	0.31	Sweets and desserts	0.20	Animal blood	0.24
Dairy products	0.27	Dairy products	0.15	Fish and Seafood	0.22
Variance explained (%)	12.8	Variance explained (%)	10.0	Variance explained (%)	8.5

^a For simplicity, only the top fifteen food items of each pattern are shown

had a multivariable HR of 1.29 (95% CI: 1.07, 1.55; P for trend < 0.01). Participants with a higher intake of the sugar rich dietary pattern had a multivariable HR of 1.22 (95% CI: 1.02, 1.46; *P* for trend = 0.03), for the same comparison. In the UK Biobank, the multivariable HRs (95% CIs) of depressive symptoms across the quartiles of the healthy dietary pattern were 1.00 (reference), 0.84 (0.72, 0.99), 0.90 (0.77, 1.00), and 0.97 (0.83, 1.14) (*P* for trend = 0.90) in the final model. For the same comparison, the multivariable HRs (95% CIs) of depressive symptoms across the quartiles of processed food dietary pattern were 1.00 (reference), 1.08 (0.92, 1.27), 1.34 (1.14, 1.58), and 1.39 (1.16, 1.68) (P for trend < 0.0001). Finally, the multivariable HRs (95% CIs) of depressive symptoms across the quartiles of processed food dietary pattern were 1.00 (reference), 0.99 (0.85, 1.15), 0.94 (0.81, 1.10), and 0.89 (0.75, 1.05) (P for trend = 0.13).

In sensitivity analyses, three dietary patterns remained similar when we include the participants who completed ≥ 1 web-based 24-h dietary assessment (Supplementary Table 1). Similar results were observed in the association between the three dietary patterns and depressive symptoms (Supplementary Table 2).

Discussion

The objective of this study was to examine the association between dietary patterns from factor analysis and depressive symptoms in participants in the TCLSIH and UK Biobank. Results indicated that in both cohorts, greater adherence to healthy dietary patterns were associated with a lower risk of depressive symptoms, and more closely to the dietary patterns rich in processed foods was significantly associated with a higher risk of depressive symptoms. However, we did not observe significant associations between meat dietary pattern and depressive symptoms risk in the UK Biobank. This investigation has further confirmed the significant association between dietary patterns and the risk of depressive symptoms.

Our results reported a positive association between dietary patterns characterized by high intakes of processed food (i.e., sugar rich and processed animal offal included in the animal food dietary pattern in the TCLSIH and processed food dietary pattern in the UK

Healthy dietary pattern		Processed food dietary pattern		Meat dietary pattern	
Food groups Factor loadings		Food groups	Factor loadings	Food groups	Factor loadings
Vegetables & potatoes	0.71	Fat & spreads	0.68	Meat & meat products	0.78
Fruits	0.70	Cereals & cereal products	0.60	Alcoholic beverages	0.50
Dairy & dairy-free products	0.41	Sugar, preserves, cakes & confectionery, snacks	0.57	Egg & egg dishes	0.26
Nuts & seeds	0.34	Non-alcoholic beverages	0.51	Vegetables & potatoes	0.15
Non-alcoholic beverages	0.31	Meat & meat products	0.20	Non-alcoholic beverages	0.09
Fish & fish dishes	0.31	Mixed-dishes	0.13	Sugar, preserves, cakes & confectionery, snacks	0.06
Egg & egg dishes	0.25	Dairy & dairy-free products	0.13	Fat & spreads	0.05
Meat substitutes	0.08	Meat substitutes	0.09	Nuts & seeds	0.02
Cereals & cereal products	0.07	Vegetables & potatoes	0.05	Fish & fish dishes	-0.11
Meat & meat products	0.02	Fruits	0.04	Mixed-dishes	-0.13
Mixed-dishes	0.02	Egg & egg dishes	0.03	Dairy & dairy-free products	-0.17
Sugar, preserves, cakes & confectionery, snacks	-0.09	Nuts & seeds	-0.04	Fruits	-0.19
Alcoholic beverages	-0.10	Alcoholic beverages	-0.05	Cereals & cereal products	-0.21
Fat & spreads	-0.15	Fish & fish dishes	-0.15	Meat substitutes	-0.53
Variance explained (%)	11.3	Variance explained (%)	10.8	Variance explained (%)	9.8

Table 2 Factor loadings for food items derived from factor analysis (principal component analysis) in the UK Biobank Cohort Study ^a

^a participants who completed at least two web-based 24-h dietary were assessment in the UK Biobank cohort

Biobank) and depressive symptoms risk. Similarly, previous cross-sectional studies have shown that dietary patterns highly loaded with processed foods were associated with a higher risk of depressive symptoms [9, 28, 29]. Moreover, three prospective cohort studies indicated a positive association between processed food patterns (derived from principal component analysis) and depressive symptoms [13, 30, 31]. In contrast, one prospective study conducted in US middle-aged and older females found no significant associations between Western dietary patterns and the new onset of depression [16]. This inconsistency may be due to differences in the study populations, sex, race, and geographical area. Indeed, a recent meta-analysis reported that when the participants were Asian and (or) less than 50 years old, the associations between healthy or Western-style dietary pattern and depression were obvious [32]. There are several plausible mechanisms underlying this association. All these dietary patterns were characterized by a poor nutritional profile with higher energy density, total fat, saturated fat, added sugar and salt, and lower amounts of dietary fiber and vitamins [33]. The low levels of the brain-derived neurotrophic factor (BDNF) induced by higher sugar consumption, which has been discussed as facilitating neurogenesis and hippocampal atrophy in depressive symptoms [34, 35]. Moreover, a previous study demonstrated the Western diet with a higher intake of processed meat, sweets, desserts, French fries, and high-fat dairy products was associated with higher levels of C-reactive protein and interleukin-6 (markers of systemic inflammation) [36]. Proinflammatory cytokines alter the production, metabolism, and transport of neurotransmitters that synergistically affect mood, including dopamine, glutamate, and serotonin [37].

We found that the traditional Chinese dietary pattern and healthy dietary pattern were associated with decreased risk of depressive symptoms. Although the Chinese traditional diet has been westernized, including reducing the intake of vegetables and whole grains and increasing the intake of refined carbohydrates, added sugar, fat, and animal source foods [38], the persistence of Chinese culinary tradition was revealed from the dietary patterns identified in the TCLSIH cohort. A cross-sectional study of Japanese people found a healthy Japanese dietary pattern with high consumption of vegetables, fruit, soy products, and mushrooms was associated with a decreased incidence of depressive symptoms [8]. Furthermore, previous meta-analysis and systematic review of observational studies both indicated that healthy dietary pattern was associated with a decreased risk of depression [32, 39]. This association might be explained by dietary fiber in vegetables and whole grains. Dietary fiber can affect the diversity and composition of intestinal microflora structure [40], which can increase serotonin concentration and reduce the production of inflammatory cytokines [41, 42]. Meanwhile,

Characteristics	All	Depressive symptoms st	P value ^b		
		No	Yes		
The TCLSIH					
No. of participants	7,094	6,105	989	-	
Age (years)	38.1 (37.9, 38.3) ^c	38.0 (37.8, 38.3)	38.4 (37.8, 39.0)	0.21	
Sex (male, %)	56.8	57.0	56.0	0.58	
BMI (kg/m ²)	24.1 (24.0, 24.2)	24.2 (24.1, 24.2)	24.0 (23.8, 24.2)	0.18	
TC (mmol/L)	4.60 (4.60, 4.70)	4.60 (4.60, 4.70)	4.60 (4.60, 4.70)	0.41	
TG (mmol/L)	1.13 (1.12, 1.15)	1.13 (1.12, 1.15)	1.14 (1.10, 1.18)	0.87	
LDL-C (mmol/L)	2.64 (2.62, 2.66)	2.64 (2.62, 2.66)	2.64 (2.59, 2.69)	0.96	
HDL-C (mmol/L)	1.33 (1.32, 1.34)	1.33 (1.32, 1.34)	1.31 (1.29, 1.34)	0.19	
FBG (mmol/L)	5.06 (5.04, 5.07)	5.06 (5.04, 5.08)	5.04 (5.00, 5.08)	0.43	
SBP (mmHg)	119.5 (119.2, 119.9)	119.5 (119.2, 119.9)	119.3 (118.3, 120.2)	0.60	
DBP (mmHq)	75.6 (75.4, 75.9)	75.7 (75.4, 76.0)	75.4 (74.7, 76.1)	0.38	
PA (MET×hour/week)	10.8 (10.5, 11.1)	11.0 (10.7, 11.4)	9.80 (9.00, 10.6)	< 0.01	
Total energy intake (kcal/day)	2283.9 (2263.6, 2304.3)	2284.6 (2262.9, 2306.4)	2279.8 (2223.6, 2337.5)	0.88	
"Vegetable rich" dietary pattern score	0.00 (-0.02, 0.02)	0.01 (-0.01, 0.04)	-0.08 (-0.14, -0.02)	< 0.01	
"Sugar rich" dietary pattern score	0.00 (-0.02, 0.02)	-0.01 (-0.03, 0.02)	0.04 (-0.03, 0.11)	0.20	
"Animal foods" dietary pattern score	0.00 (-0.02, 0.02)	-0.02 (-0.05, 0.00)	0.14 (0.07, 0.21)	< 0.0001	
Smoking status (%)	,	()		0.01	
Current smoker	17.9	17.3	21.3		
Fx-smoker	5.21	5.29	4.70		
Non-smoker	76.9	77.4	74.0		
Drinking status (%)	, 0.5		7.10	0.52	
Everyday	3 10	2 98	3 80	0.02	
Sometime	61 3	61.2	61.6		
Ex-drinker	8.98	9.06	852		
Non-drinker	26.7	26.7	26.1		
Married (%)	85.2	85.2	85.1	0.93	
Living alone (%)	7.45	7.28	8.47	0.55	
Education level (college or higher %)	80.9	82.1	73.7	< 0.0001	
Occupation (%)	00.9	02.1	75.7	0.48	
Managers	51.8	52.0	50.3	0.10	
Professionals	18.5	18.5	183		
Other	20.8	20.5	31 /		
Visiting friends (%)	55.6	567	50.7	< 0.001	
Household income $(> 10,000$ Yuan %)	16.8	18.4 18.1	36.8	< 0.001	
Individual history of disease (%)	40.0	40.4	50.0	< 0.0001	
Hypertension	21.2	21.2	21.2	0 00	
Hyperlinidemia	13.8	/3.5	21.2 /1.0	0.55	
Diabatas	45.8	43.5	41.0	0.15	
Eamily history of disease (%)	ч.) 5	UU	7.20	0.42	
	22.7	22.6	244	0.62	
Hypertension	55.7	55.0	54.4	0.02	
Hypertension	0.40	0.47	0.61	0.20	
Disheter	0.49	0.47	21.4	0.47	
	29.1	29.0	21.4	0.23	
No of participants	06.810	05 507	1 303	_	
	90,010 EAQ (EAQ EAQ)	33,JUI EAQ (EAQ EAQ)		-	
Aye (years)	J4.0 (J4.0, J4.9)	24.0 (24.0, 24.9)	J4.4 (J4.U, J4.8)	0.03	

Table 3 Baseline participant characteristics by status of depressive symptoms ^a

Table 3 (continued)

Characteristics	All	Depressive symptoms status		P value ^b	
		No	Yes		
BMI (kg/m ²)	26.3 (26.2, 26.3)	26.2 (26.2, 26.3)	27.9 (27.6, 28.2)	< 0.0001	
SBP (mmHg)	132.9 (132.8, 133.0)	132.9 (132.8, 133.1)	131.1 (130.0, 132.3)	< 0.01	
DBP (mmHg)	79.8 (79.7, 79.9)	79.8 (79.7, 79.9)	79.7 (79.0, 80.3)	0.65	
Townson depretive index	-1.62 (-1.64, -1.60)	-1.63 (-1.64, -1.61)	-1.11 (-1.28, -0.94)	< 0.0001	
PA (MET×hour/week)	26.0 (25.8, 26.2)	26.1 (25.9, 26.2)	22.0 (20.5, 23.6)	< 0.0001	
Total energy intake (kcal/day)	2039.2 (2035.8, 2042.5)	2039.6 (2036.2, 2042.9)	2009.8 (1979.9, 2040.1)	0.06	
"Healthy" dietary pattern score	0.00 (-0.01, 0.01)	0.00 (-0.01, 0.01)	0.05 (-0.01, 0.11)	0.13	
"Processed food" dietary pattern score	0.00 (-0.01, 0.01)	0.00 (-0.01, 0.00)	0.10 (0.04, 0.16)	< 0.001	
"Meet" dietary pattern score	0.00 (-0.01, 0.01)	0.00 (-0.01, 0.01)	-0.04 (-0.10, 0.02)	0.17	
Smoking status (%)				< 0.0001	
Current smoker	6.87	6.79	12.2		
Ex-smoker	34.6	34.5	37.2		
Non-smoker	58.6	58.7	50.6		
Drinking status (%)				< 0.0001	
Current drinker	94.5	94.6	89.3		
Ex-drinker	2.70	2.64	6.68		
Non-drinker	2.79	2.77	3.99		
Living alone (%)	17.6	17.5	23.9	< 0.0001	
Education level (college or higher, %)	47.6	47.7	40.2	< 0.0001	
Visiting friends (≥once a week, %)	74.9	74.9	76.1	0.34	
Individual history of disease (%)					
Hypertension	21.1	21.0	28.1	< 0.0001	
Diabetes	2.84	2.80	5.53	< 0.0001	
Family history of disease (%)					
CVD	54.0	53.9	58.9	< 0.001	
Hypertension	44.2	44.2	46.2	0.14	
Diabetes	16.9	16.9	18.7	0.07	

^a Continuous variables are expressed as means (± standard deviation, SD) and categorical variables are expressed as percentages. BMI, body mass index; CVD, cardiovascular disease; DBP, diastolic blood pressure; FBG, fasting blood glucose; HDL-C, high-density lipoprotein cholesterol; LDL-C, low-density lipoprotein cholesterol; MET, metabolic equivalent; PA, physical activity; SBP, systolic blood pressure; TC, total cholesterol; TG, triglycerides

 b T test or χ^{2} test

^c Geometric mean (95% confidence interval) (all such values)

the potential protective effect of the traditional Chinese dietary pattern could also come from the high content of antioxidants in vegetables such as vitamin *C*, vitamin *E*, and other carotenoids compounds, as some studies have shown higher antioxidant levels to be associated with a lower risk of depressive symptoms [43]. Interestingly, participants with the highest quartiles of healthy dietary pattern scores were not with a lower risk of depressive symptoms. That might be partly due to a higher intake of starchy vegetables which may not confer the same health benefits as other fruits and vegetables [44].

However, in the UK Biobank cohort, the higher consumption of meat dietary pattern was not associated with a high risk of depressive symptoms. Previous studies have some controversial reports on meat intake and depressive symptoms [30, 45]. The uncertain association may be due to the effect of white, red, and processed meat on depression may differ from each other [46]. In the present study, the various varieties of meat could only be regarded as a whole. That means the results of this study are subject to the combined effect of different varieties of meat. Furthermore, most of the participants in the UK Biobank cohort prefer to remove the fat from meat and remove the skin from poultry before cooking, which could also reduce the detrimental effect of meat intake on depression.

Strengths of this study include its large sample size of TCLSIH and UK Biobank participants, and a dietary assessment tool that allowed for detailed estimation of food and nutrient intakes as well as overall dietary

Table 4 Association between dietary patterns and risk of depressive symptoms in the TCLSIH and UK Biobank Cohort Study

	Quartiles of dietary pattern scores			P for trend ^a	
	Q1	Q2	Q3	Q4	
The TCLSIH					
Traditional Chinese dietary pattern					
No. of depressive symptoms	278	249	230	232	
Person years	4,255	4,375	4,392	4,389	
Incidence per 1000 person years	65.33	56.91	52.37	52.86	
Model 1	1.00 (reference)	0.86 (0.73, 1.02) ^b	0.79 (0.66, 0.94)	0.79 (0.67, 0.94)	< 0.01
Model 2	1.00 (reference)	0.85 (0.72, 1.01)	0.78 (0.65, 0.94)	0.71 (0.57, 0.88)	< 0.001
Model 3	1.00 (reference)	0.85 (0.71, 1.01)	0.78 (0.65, 0.94)	0.71 (0.57, 0.88)	< 0.001
Processed animal offal included anir	mal food dietary pattern				
No. of depressive symptoms	248	215	221	305	
Person years	4,365	4,391	4,360	4,295	
Incidence per 1000 person years	56.82	48.96	50.69	71.01	
Model 1	1.00 (reference)	0.88 (0.73, 1.06)	0.93 (0.77, 1.12)	1.32 (1.11, 1.56)	< 0.001
Model 2	1.00 (reference)	0.87 (0.72, 1.05)	0.93 (0.77, 1.12)	1.29 (1.07, 1.54)	< 0.01
Model 3	1.00 (reference)	0.87 (0.73, 1.05)	0.93 (0.77, 1.12)	1.29 (1.07, 1.55)	< 0.01
Sugar rich dietary pattern					
No. of depressive symptoms	232	241	250	266	
Person years	4,420	4,298	4,418	4,276	
Incidence per 1000 person years	52.49	56.07	56.59	62.21	
Model 1	1.00 (reference)	1.08 (0.90, 1.29)	1.08 (0.91, 1.30)	1.21 (1.01, 1.44)	0.04
Model 2	1.00 (reference)	1.09 (0.91, 1.31)	1.11 (0.93, 1.33)	1.22 (1.02, 1.46)	0.03
Model 3	1.00 (reference)	1.09 (0.91, 1.31)	1.11 (0.93, 1.33)	1.22 (1.02, 1.46)	0.03
The UK Biobank					
Healthy dietary pattern					
No. of depressive symptoms	349	289	310	355	
Person years	177,189	177,755	177,974	177,013	
Incidence per 1000 person years	1.97	1.63	1.74	2.01	
Model 1	1.00 (reference)	0.81 (0.69, 0.95)	0.85 (0.73, 1.00)	0.95 (0.81, 1.10)	0.65
Model 2	1.00 (reference)	0.85 (0.72, 0.99)	0.90 (0.77, 1.00)	0.98 (0.84, 1.15)	0.99
Model 3	1.00 (reference)	0.84 (0.72, 0.99)	0.90 (0.77, 1.00)	0.97 (0.83, 1.14)	0.90
Processed food dietary pattern					
No. of depressive symptoms	295	301	361	346	
Person years	177,848	177,906	177,636	176,541	
Incidence per 1000 person years	1.66	1.69	2.03	1.96	
Model 1	1.00 (reference)	1.04 (0.89, 1.22)	1.30 (1.11, 1.51)	1.35 (1.15, 1.58)	< 0.0001
Model 2	1.00 (reference)	1.09 (0.92, 1.28)	1.36 (1.16, 1.60)	1.43 (1.18, 1.72)	< 0.0001
Model 3	1.00 (reference)	1.08 (0.92, 1.27)	1.34 (1.14, 1.58)	1.39 (1.16, 1.68)	< 0.0001
Meat dietary pattern					
No. of depressive symptoms	336	335	324	308	
Person years	177,286	177,802	177,632	177,211	
Incidence per 1000 person years	1.90	1.88	1.82	1.74	
Model 1	1.00 (reference)	0.96 (0.83, 1.12)	0.92 (0.79, 1.08)	0.91 (0.78, 1.07)	0.21
Model 2	1.00 (reference)	0.99 (0.85, 1.15)	0.95 (0.81, 1.10)	0.90 (0.76, 1.05)	0.16
Model 3	1.00 (reference)	0.99 (0.85, 1.15)	0.94 (0.81, 1.10)	0.89 (0.75, 1.05)	0.13

^a Obtained by using multivariable Cox regression model

^b Hazard ratios (95% confidence interval) (all such values)

Model 1 was adjusted for age, sex, and body mass index

Model 2 was additionally adjusted for smoking status, alcohol drinking status, married (only in TCLSIH cohort), education level, occupation (only in TCLSIH cohort), visiting friends, living alone, household income (only in TCLSIH cohort), physical activity, total energy intake, Townson depressive index (only in UK Biobank) Model 3 was additionally adjusted for family history of disease (including cardiovascular disease, hypertension, hyperlipidemia (only in TCLSIH cohort), and diabetes), hypertension, hyperlipidemia (only in TCLSIH cohort), diabetes patterns. Furthermore, these data allowed us to investigate the association between dietary patterns and depressive symptoms with diet collected before the onset of the disease, which can reduce reverse causal associations compared to cross-sectional studies. We also adjusted for a variety of covariables and conducted sensitivity analyses to confirm the robustness of derived dietary patterns and their association with depressive symptoms. However, a few limitations are notable in the present study. First, the statistical analysis used to derive dietary patterns involves several arbitrary decisions, including the construction of the food groups and the method of rotation. However, similar dietary patterns were identified when including individual food items instead of food groups in our previous study [47]. In addition, three dietary patterns remained similar when we included the participants who completed ≥ 1 WebQ dietary assessment in the UK Biobank study. Second, the measurement of depressive symptoms in the TCLSIH study depends on SDS rather than conducting diagnostic psychiatric interviews. However, the sensitivity and specificity of the SDS score for the assessment of depressive symptoms compared to the Diagnostic and Statistical Manual of Mental Disorders (fourth edition), were 83.6% and 96.4%, respectively. Third, dietary intake was assessed using the 100-item FFQ or WebQ dietary assessment which cannot include all the food items and could be subject to recall bias [48]. Therefore, the present dietary pattern may not reflect the real situation of the study population. Fourth, the study participants were mainly Chinese adults and British whites, and our findings may only apply to the population with similar demographics. Finally, even though many confounding factors have been taken into account, some potential residual factors are still unavoidable to confound the observed relationship.

Conclusions

Based on two large China and UK cohorts, following a traditional Chinese dietary pattern or healthy dietary pattern was associated with a lower incidence of depressive symptoms, while a higher intake of a diet rich in processed food was associated with an increased risk of depressive symptoms. Furthermore, following a meat dietary pattern was not associated with depressive symptoms in the UK. These findings suggest that dietary modification may be a potential target for the prevention of depressive disorders. Further research is urgently required to confirm the causal association of dietary patterns with the risk of depressive symptoms, and they need to be confirmed in other populations and settings.

Abbreviations

	Drain derived neurotraphic factor
DUNF	brain-derived neurotrophic factor
BMI	Body mass index
BP	Blood pressure
CVD	Cardiovascular disease
CI	Confidence interval
FBG	Fasting blood glucose
FFQ	Food frequency questionnaire
HDL-C	High-density lipoprotein cholesterol
HR	Hazard ratio
ICD	International Classification of Diseases
LDL-C	Low-density lipoprotein cholesterol
MET	Metabolic equivalent
SDS	Self-Rating Depression Scale
TC	Total cholesterol
TCLSIH	Tianjin Chronic Low-grade Systemic Inflammation and Health

TG Triglycerides

Supplementary Information

The online version contains supplementary material available at https://doi. org/10.1186/s12966-023-01461-x.

Additional file 1.

Acknowledgements

The authors gratefully acknowledge all the people that have made this study.

Authors' contributions

H.W. analyzed the data and wrote the paper. H.W., Y.G., G.M., Q.Z., L.L., H.W., S.Z., T.Z., X.W., J.Z., S.S., X.W., M.Z., Q.J., H.C., T.H., and K.S. conducted research. K.N. designed the research and had primary responsibility for the final content. All authors had access to the study data and reviewed and approved the final manuscript.

Funding

This study was supported by grants from the Study of Diet and Nutrition Assessment and Intervention Technology (No. 2020YFC2006300) from Active Health and Aging Technologic Solutions Major Project of National Key R&D Program——Study on Intervention Strategies of Main Nutrition Problems in China (No. 2020YFC2006305), National Natural Science Foundation of China (Nos. 81941024, 81872611, 82103837 and 81903315), Tianjin Major Public Health Science and Technology Project (No. 21ZXGWSY00090), National Health Commission of Chinase Institute of Food Science and Technology Foundation of Chinese Institute of Food Science and Technology (No. 2019–12), 2014 and 2016 Chinese Nutrition Society (CNS) Nutrition Research Foundation—DSM Research Fund (Nos. 2016–046, 2014–071 and 2016–023), China. The funding body had no role in the study design; collection, analysis, and interpretation of data; or writing of the manuscript.

Availability of data and materials

TCLSIH data cannot be made publicly available because public availability would compromise participant privacy. For data access, researchers can contact the School of Public Health, Tianjin University of Traditional Chinese Medicine, Tianjin, China (E-mail address: nkj0809@gmail.com). UK Biobank data are available in a public, open-access repository. This research has been conducted using the UK Biobank Resource under Application Number 44430. The UK Biobank data are available on the application to the UK Biobank (www.ukbiobank.ac.uk/).

Declarations

Ethics approval and consent to participate

Ethics approval was provided by the Institutional Review Board of Tianjin Medical University (reference number: TMUh-MEC 201430). The UK Biobank Study was approved by the North West Multi-centre Research Ethics Committee (REC reference for UK Biobank 11/NW/0274). Besides, all the participants have written informed consent.

Consent for publication

Not applicable.

Competing interests

Not applicable.

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Received: 7 July 2022 Accepted: 30 April 2023 Published online: 20 June 2023

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