Dietary patterns and cardiometabolic risks in diverse less-developed ethnic minority regions: results from the China Multi-Ethnic Cohort (CMEC) Study

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A B S T R A C T

Background In Western developed countries, food-based dietary patterns have been associated with the risk of cardiometabolic diseases, but little is known about such associations in less developed ethnic minority regions (LEMRs), where the cardiometabolic disease burden is growing rapidly and food patterns differ substantially.

Methods Between May 2018 and September 2019, we recruited 99556 participants aged 30-79 years from the China Multi-Ethnic Cohort (CMEC) Study. We measured habitual dietary intake with validated food frequency questionnaire (FFQ) and then calculated dietary pattern scores for two of the most studied a priori dietary patterns, i.e., Dietary Approaches to Stop Hypertension (DASH) and alternative Mediterranean (aMED) style diets, and three a posteriori dietary patterns. Four cardiometabolic risks, including hypertension, diabetes, dyslipidaemia and metabolic syndrome (MetS), were newly diagnosed by medical examination and blood tests. We estimated adjusted odds ratios (OR) relating various dietary pattern scores to cardiometabolic risks using marginal structural models under the guidance of directed acyclic graphs. For the above associations, we further calculated the proportion mediated by overweight (PM) using regression-based mediation analysis for better public health implications.

Findings The final study sample consisted of 68384 participants. Among them, we newly diagnosed 12803 hypertension, 3527 diabetes, 16342 hyperlipidaemia, and 8198 MetS cases. Overall, all 5 dietary patterns showed considerable associations with risks of hypertension and MetS. Comparing the highest with the lowest quintiles, the DASH score showed the strongest inverse associations with risks of hypertension (OR=0.74, 95% CI:0.70-0.79; PM=10%) and MetS (OR=0.79, 95% CI:0.74-0.85; PM=35%); conversely, scores of the localized a posteriori Yunnan-Guizhou plateau dietary pattern in LEMRs showed the strongest positive associations with risks of hypertension (OR=1.44, 95% CI:1.35-1.52; PM=10%) and MetS (OR=1.35, 95% CI:1.26-1.46; PM=33%), with all P values for trend <0.001. These associations were consistent in various subgroups defined by sex, age, smoking and physical activity, but with magnitudes varying across different regions.

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that differed substantially across different ethnic regions and urbanicity. By investigating the single-component effects of dietary patterns, the dairy intake component contributed a major proportion to the beneficial effects of DASH (41.9% for hypertension and 100.5% for MetS).

Interpretation Substantial socioeconomic status and ethnic disparities in diet quality and related cardiometabolic risks were seen in LEMRs, with hypertension being the top diet-related cardiometabolic risk. Our findings support that DASH provides superior dietary guidance compared to aMED for reducing cardiometabolic risks in LEMRs. In particular, the dairy intake encouraged by DASH may produce considerable beneficial effects.

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Introduction

Cardiometabolic diseases are the top-ranked causes of morbidity and mortality worldwide. In 2019, it was estimated that over 25% of the global disability-adjusted life-years (DALYS) in the elderly population are attributable to three major cardiometabolic diseases: ischaemic heart disease, stroke and diabetes.\(^1\) Over past decades, cardiometabolic disease mortality has experienced a steady decline in high-income countries (HICs)\(^2\)-\(^4\) but has increased rapidly in low- and middle-income countries (LMICs)\(^2\)-\(^4\) as well as in populations of low socioeconomic status (SES) and racial/ethnic minorities.\(^5\)-\(^12\) Reducing SES and racial/ethnic disparities in cardiometabolic diseases has therefore become a rising public health concern globally and a top priority to achieve United Nations Sustainable Development Goals (SDGs) target 3.4,\(^13\),\(^14\)

An unhealthy diet is the leading modifiable risk factor for cardiometabolic diseases.\(^15\),\(^16\). In particular, both randomized trials and long-term cohort studies suggest that healthy food-based dietary patterns produce significant benefits on cardiometabolic diseases\(^17\)-\(^19\) but that little benefit was identified from controlling single isolated nutrients.\(^20\)-\(^22\) Dietary patterns also have the advantage of public health implications because they can facilitate dietary guidance and lessen industry manipulation.\(^23\),\(^24\) For such reasons, healthful food-based dietary patterns such as Dietary Approaches to Stop Hypertension (DASH) and Mediterranean (MED)-style diets have been recommended by the U.S. Department of Agriculture and American Heart Association and used worldwide to reduce the risk of cardiometabolic diseases.\(^24\)-\(^26\) Nevertheless, almost all well-known dietary pattern guidance has been developed based on Western-like diets from developed countries. An emerging question then is whether Western dietary guidance can be extrapolated to other populations, particularly racial/ethnic minority groups in less-developed regions, for whom the cardiometabolic disease burden is growing rapidly and food patterns differ substantially.\(^27\)-\(^29\) Besides, as the above populations have been rarely studied before, another crucial question is what insights we can gain from their disparate food patterns, which may also help for cardiometabolic disease prevention. However, reliable evidence from large-scale epidemiological studies on this topic is scarce.

The China Multi-Ethnic Cohort (CMEC) Study is a large-scale epidemiological study undertaken in Southwest China (an area of 2.3 million square kilometres), covering the Qinghai-Tibet Plateau, Yunnan-Guizhou Plateau and Sichuan Basin, with great diversity in SES, ethnicity, cardiometabolic disease burden, habitual diet, living environment, etc.\(^30\) Nearly 0.1 million participants from seven ethnic groups with comprehensive information were enrolled in the CMEC study. Overall, CMEC presents an ideal and unique opportunity to finely characterize the relationship between dietary patterns and cardiometabolic disease risks in the setting of diverse less-developed ethnic minority regions (LEMRs).

In this cross-sectional analysis of CMEC baseline data, we aimed to assess associations of two of the most studied a priori dietary patterns from Western developed countries, i.e., DASH and alternative MED diets, as well as three a posteriori dietary patterns derived from CMEC dietary data with newly diagnosed cardiometabolic risks (hypertension, diabetes, dyslipidaemia and metabolic syndrome). Given that increasing evidence suggests that associations between diet and cardiometabolic risks might be partly mediated by overweight,\(^31\)-\(^33\) we further aimed to examine how such associations between dietary patterns and cardiometabolic risks might be mediated by overweight to obtain better public health implications.

Methods

Study population

The CMEC study is an ongoing community-based prospective cohort study undertaken in Southwest China, where is the major ethnic area in China and home to 56 ethnic groups. Ethnic characteristics, SES, population size and disease patterns were given special consideration when selecting the study population and survey sites. A detailed description of the study design, sampling strategy and baseline characteristics has been published elsewhere.\(^30\) In brief, a total of 99556 participants from six ethnic minority groups (Tibetan, Yi, Miao, Bai, Bouyei and Dong) as well as the majority Han group were recruited between May 2018 and September 2019. The baseline survey consisted of a tablet-based electronic questionnaire via face-to-face interviews, anthropometric measurements, thorough medical examinations, and blood and urine tests. All the participants provided written informed consent prior to the data collection, and the study was approved by the Sichuan University Medical Ethical Review Board and local ethics committee at each participating site.

In the present study, we focused on adults aged 30-79 years with complete and plausible diet- and outcome-related data, total energy intakes and body mass index (BMI). We excluded 22618 participants who self-reported physicians diagnosed hypertension, diabetes, hyperlipidaemia, coronary heart disease, or stroke at the baseline survey to eliminate potential reverse causality. The final sample consisted of 68834 participants. More details are provided in the Appendix text 1s.

Assessment of dietary intakes

For the baseline survey, habitual diets were assessed using quantitative food frequency questionnaire (FFQ) (see the full text in the Appendix text 2s). For each food group, participants were required to report the quantity (average grams per meal according to standard serving size moulds) and frequency (four frequency
categories ranged from how many times per day to year) that they consumed during the past 12 months. We also asked information about alcohol, tea, Sugar Sweetened Beverages (SSBs), cooking oil and salt in separate sections. In particular, daily alcohol consumption was calculated as grams of pure alcohol according to alcohol type, amount drunk and frequency. From the above FFQ, we estimated the total daily energy intake according to the China food exchange lists and the 2018 China food composition tables (see more details in the Appendix text 3s,35,36).

We conducted the repeated FFQ and 24-hour dietary recall (24 HDRs) to assess the reproducibility and validity of the baseline FFQ. Regarding reproducibility, intraclass correlation coefficients (ICC) for food groups ranged from 0.15 for fresh vegetables to 0.67 for alcohol. Regarding validity, de-attenuated Spearman rank correlation coefficients for food groups ranged from 0.10 for soybean products to 0.66 for rice. More details are provided in the Appendix text 4s.

**Assessment of dietary patterns**

To better capture the dietary features of our study population, we scored all participants according to their adherence to specific dietary patterns, including the two of the most studied a priori dietary patterns (i.e., DASH- and MED-style diets) relating to cardiometabolic diseases, as well as three a posteriori dietary patterns derived from CMEC data.

To represent adherence to a DASH-style diet, we used a modified DASH score with nonfat and low-fat dairy replaced by full-fat dairy products, given that the consumption of nonfat and low-fat dairy was extremely low in our study population. Compared to the original DASH diet, the modified DASH diet was shown by a randomized controlled trial to be more effective at reducing cardiometabolic risks. In addition, we excluded the food group component of Sugar Sweetened Beverages (SSBs) because regular consumption of SSBs in our study population was only 7.2%. To represent adherence to a MED-style diet, we used an alternative Mediterranean diet (aMED) score, which is an adapted version of the traditional MED for the non-Greek population. We also eliminated the food group component of nuts due to the lack of a separate food group for nut consumption. For each of food group components of DASH or aMED, we categorized the food group consumption into quintiles and scored all participants from 1 to 5 according to their intake ranking, and then obtained the total score by summing up the component scores. More details are provided in the Appendix text 5s.

To determine the major a posteriori dietary patterns, we used principal component factor analysis (PCFA) with varimax rotation based on the same food group data transformed to z scores. Three major dietary patterns were identified via comprehensive considerations of eigenvalues, variance explained, scree plot and interpretability (see more details in the Appendix text 6s). For each dietary pattern, factor scores were calculated for all participants by summing up the standardized intakes of food groups weighted by their factor loadings. The robustness of the characteristics of three identified dietary patterns were assessed by a split-sample validation (see more details in the Appendix text 7s).

**Assessment of outcomes**

The outcomes in this study were newly diagnosed cardiometabolic risks, including hypertension, diabetes, dyslipidaemia and metabolic syndrome (MetS). We identified all cardiometabolic risks based on objective indicators from medical examinations or blood tests at the baseline survey. More detailed information about assessment of cardiometabolic outcomes is provided in the Appendix text 8s.

**Assessment of covariates**

We obtained covariate information from the baseline questionnaire. To guide the selection of potential confounders, we constructed directed acyclic graphs (DAGs) under the protocol of “Evidence Synthesis for Constructing Directed Acyclic Graphs” (ES- DAGs), which combined evidence synthesis strategies and causal inference principles. We further performed independent tests to continuously modify the proposed DAGs until the implied conditional independences were satisfied. See more details in the Appendix text 9s and e-table. On the basis of these causal diagrams and backdoor criteria, we adjusted the final models for sex, age, urbanicity, ethnicity, marital status, highest education attained, annual household income, occupation, regular smoking, physical activity in hours of metabolic equivalent tasks per day (METs-h/day), total energy intake (kcal per day), regular intake of sweeten beverage, regular intake of dietary supplements, regular intake of spicy food, regular intake of pepper food, insomnia symptoms, depressive symptom, anxiety symptom, menopause status for women, and family history of cardiometabolic diseases.

**Statistical analysis**

To assess ethnic and regional variations in dietary patterns, we compared the median (25th, 75th percentiles) of five dietary pattern scores across ethnic regions. We categorized all dietary pattern scores into quintiles for the entire study population. Baseline characteristics are described as the median (25th, 75th percentiles) for continuous variables and percentage for categorical variables according to quintiles of dietary pattern scores.

We employed marginal structural models which combined logistic regression with the inverse probability of exposure weighting (IPEW) to estimate associations between the five dietary pattern scores (quintiles of dietary pattern scores were modelled as categorical variable with five levels) and four cardiometabolic risks separately, with the lowest fifth of the dietary pattern score as the reference group. To determine the preferable weighting method, we adopted six weighting methods in our primary analysis, with the quintiles of dietary pattern scores as the dependent variable and confounders decided by the ESC-DAGs as the independent variables, and then assessed their balances of confounders among different exposure groups. In our final model, we used the entropy balancing weighting method due to the optimal balance of confounders. See the detailed results in the Appendix text 10s. To evaluate the linear trend across quintiles, we assigned a median value to each quintile of dietary scores and then modelled the median score as a continuous variable. To facilitate interpretations of the effects of overall dietary patterns, we carried out single-component and single group analyses. For DASH and aMED, we evaluated the association of each of the dietary components with cardiometabolic risks by eliminating one component at a time from the overall score separately.

We also assessed the association of each of the single food groups in our FFQ with cardiometabolic risks. To examine the mediation effects of overweight (BMI <24 OR ≥24 kg/m²), regression-based mediation analyses were used to decompose the total effects of dietary patterns on cardiometabolic risks into natural direct and indirect effects through overweight and to calculate the proportion of mediation accordingly. Due to the built-in quality control in the tablet-based questionnaire and stringent data audit, the missing proportion in this study was very low (<5% except for oil and salt consumption). Most missing values were generated from unverifiable outliers after audio review. For missing values of food groups, we performed multiple imputation (with 5 imputations) by the chained equations method. To simultaneously account for the uncertainty of estimations of both weighting and...
exposure-outcome associations, a bootstrap-based method with 1000 replications was used to obtain 95% confidence intervals (CIs). To examine potential effect modifiers, we conducted stratification analysis among predefined subgroups, including sex, age, regular smoking, physical activity, urbanicity and ethnic regions. Heterogeneity among different strata was assessed using the chi square ($\chi^2$) test. To assess the robustness of our findings, we also performed several sensitivity analyses. First, we used a more stringent exclusion criteria by excluding self-reported physicians diagnosed hepatic and gastrointestinal diseases. Second, we used all identified cardiometabolic risks (self-reported plus newly identified) as outcomes to examine the magnitude of potential reverse causality. Third, we adopted the conventional covariate adjustment approach to estimate the associations. Fourth, we alternatively adjusted for BMI as a confounder rather than a mediator. Fifth, we ran a minimally adjusted analysis, without justification for spicy food, pepper food, insomnia, depression, and anxiety. Sixth, we alternatively calculated the DASH score with the inclusion of SSBSs. Last, we ran a complete case analysis instead of the multiple imputation approach.

All analyses were performed with R Project for Statistical Computing version 4.0.2 (Vienna, Austria). The maps displayed in the results were produced by ArcGIS Desktop version 10.2.1 (authorization number: EFL734321752).

Role of the funding source

The funder of the study had no role in study design, data collection, data analysis, data interpretation, or writing of the report. XX, YJ, and XZ had access to all data and had final responsibility for the decision to submit for publication.

Results

Characteristics of dietary patterns

Figure 1 shows the geographic and ethnic variations of various dietary pattern scores. Regarding Western-like diet-based a priori dietary patterns, no substantial difference in median scores was found across different regions and ethnic groups, with DASH scores ranging from 17 (Dong in Guizhou) to 22 (Han in Basin) and aMED scores ranging from 21 (Tibetans in Aba) to 26 (Han in Basin) (figure 1B). In contrast, a posteriori dietary patterns showed a better ability to discriminate different ethnic groups, with various ethnic groups clustered into three geography-related a posteriori dietary patterns. The first one predominated by the Han majority population in the more developed Sichuan Basin (with two mega-cities located in this region) was characterized by relatively high intakes of fish/sea food, poultry, eggs, dairy products, fresh fruits, fresh vegetables and vegetable oil, implying a modern and balanced dietary pattern: the Sichuan Basin dietary pattern. The second one predominated by various ethnic minority populations in the less developed Yunnan-Guizhou Plateau was characterized by relatively high intakes of animal oil, rice, salt, preserved vegetables, alcohol and soybean products and low intakes of dairy products and coarse grain, which suggested a poor and agricultural dietary pattern: the Sichuan Basin dietary pattern. The third one predominated by the Tibetan population in the less developed Qinghai-Tibet Plateau was characterized by relatively high intakes of coarse grain, wheat products, tubers and tea, indicating a featured high-altitude Tibetan dietary pattern: the Qinghai-Tibet Plateau dietary pattern (figure 1A and 1C). More detailed information in table format on the comparisons of dietary scores across ethnic regions is provided in the Appendix text 11s.

Baseline characteristics according to the lowest and highest quintiles of the various dietary scores are reported in Table 1. Among the 68834 participants included in this study, the median age was 49.5 (42, 56) years, 62% of the participants were women, 35% of the participants were urban residents, and 53% of the participants were ethnic minority populations. Participants with higher adherence to DASH, aMED and Sichuan Basin dietary patterns shared similar characteristics. They tended to have higher education and SES levels as well as healthier lifestyles, and they were less likely to report mental disorders but more likely to report a family history of cardiometabolic diseases. In contrast, participants with higher adherence to the Yunnan-Guizhou Plateau or Qinghai-Tibet Plateau dietary patterns generally tended to have lower education and SES levels as well as unhealthier lifestyles.

Associations of dietary patterns with cardiometabolic risks

In the baseline survey of the CMEC study, we newly identified cases of 12803 hypertension, 3527 of diabetes, 16342 of hyperlipidaemia and 8198 of MetS. Figure 2 displays the estimated associations between dietary patterns and cardiometabolic risks after adjusting for potential confounders. Overall, dietary patterns showed considerable associations with risks of hypertension and MetS but only borderline associations with risks of diabetes and hyperlipidaemia. Comparing the highest with lowest quintiles, the DASH score showed the strongest inverse associations with risks of hypertension (OR=0.74, 95% CI=0.69-0.80) and MetS (OR=0.79, 95% CI=0.73-0.87), whereas the score of a posteriori Yunnan-Guizhou plateau dietary pattern showed the strongest positive associations with risks of hypertension (OR=1.44, 95% CI=1.27-1.63) and MetS (OR=1.35, 95% CI=1.15-1.57), with all P values for trend <0.001. For other dietary patterns, the Sichuan Basin showed similar but weaker associations with various cardiometabolic risks as DASH; a similar situation was also found between the Qinghai-Tibet and Yunnan-Guizhou Plateau dietary patterns. See the data presented in Figure 2 in the table format in the Appendix text 12s.

Although DASH and aMED shared some common food group components, aMED had notably different impacts on cardiometabolic risks. By comparing the highest with the lowest quintiles, aMED was associated with a 14% lower risk of hypertension (OR=0.86, 95% CI=0.80-0.93) but a 14% higher risk of MetS (OR=1.14, 95% CI=1.05-1.25) (figure 2). The discrepant results between DASH and aMED can be well confirmed by single-component analyses (Appendix text 13s: tables 19-22s). Among the discrepant food group components, the dairy product components included in DASH (but not in aMED) contributed a majority of the beneficial effects of DASH on hypertension (41.9%) and MetS (100.5%). In contrast, the monounsaturated fatty acids: saturated fatty acids (MUFA: SFA) ratio included in aMED (but not in DASH) had harmful effects, contributing a major proportion to the positive associations of aMED with hypertension (62.7%) and MetS (83.1%). Single food group analyses also showed similar results, especially for the strongest inverse association between dairy products and cardiometabolic risks (Appendix text 13s: tables 23s).

For simplicity, only associations of DASH and Yunnan-Guizhou plateau dietary patterns with hypertension and MetS are reported in the following results. In stratification analysis, the inverse associations of DASH and the positive associations of the Yunnan-Guizhou plateau dietary pattern with hypertension and MetS were consistent across various subgroups, with the only exceptions being ethnic region and urbanicity (figure 3). Both dietary patterns showed stronger associations with hypertension among the Han majority in Sichuan Basin compared to the ethnic minorities in other regions. For MetS, dietary patterns were prone to exhibit a risker association among the ethnic minorities on the Qinghai-Tibet Plateau compared to others, probably due to the extraordinarily high prevalence of central obesity in the Tibetan population. The stratified results of urbanicity highly agreed with those of ethnic regions because nearly all of the ethnic minority populations lived in rural areas. By decomposing the total
Figure 1. Geographic and ethnic variations of various dietary patterns among various ethnic groups. (A) Geographic locations and terrains of study sites. The administrative boundary data with scale of 1:4 million was obtained from national fundamental geoinformation database. The digital elevation data with 30 meters resolution used in terrain map was obtained from the advanced spaceborne thermal emission and reflection radiometer global digital elevation Model. (B) Comparisons of a priori dietary pattern scores (includes DASH and aMED) among various ethnic groups. (C) Characteristics of three a posteriori dietary patterns and related comparisons among various ethnic groups. Boxplot based on median (25th, 75th percentiles) was used to visually compare the distribution of dietary scores across ethnic regions.
Table 1
Baseline characteristics in the CMEC Study, according to quintiles of various dietary pattern scores. 

| Characteristic                      | No. of participants | Overall | DASH | aMED | Sichuan Basin | Yunnan-Guizhou Plateau | Qinghai-Tibet Plateau |
|-------------------------------------|---------------------|---------|------|------|----------------|-------------------------|-----------------------|
|                                     |                     |         | Q1   | Q5   | Q1            | Q5                      | Q1            | Q5            | Q1  | Q5  | Q1  | Q5  |
| Dietary score                       | 68834               |         | 15   | 26   | 19            | 30                      | -1.1          | 1.4           | -1.2 | 1.3 | -1.0 | 1.2 |
| Age (yr)                            | 68834               | 49      | 50   | 47   | 50            | 48                      | 51            | 47            | 46  | 51  | 49   | 49  |
| Female sex (%)                      | 68834               | 62      | 54   | 70   | 53            | 69                      | 66            | 57            | 75  | 47  | 75   | 50  |
| Ethnic group (%)                    |                     | 32116   | 47   | 36   | 62            | 29                      | 15            | 79            | 62  | 31  | 42   | 35  |
| Sichuan Basin                       |                     | 29924   | 43   | 57   | 30            | 51                      | 63            | 18            | 11  | 68  | 56   | 32  |
| Yunnan-Guizhou Plateau              |                     | 6794    | 10   | 7    | 8             | 19                      | 23            | 3             | 27  | 2   | 1    | 34  |
| Qinghai-Tibet Plateau              |                     | 68834   | 35   | 26   | 49            | 22                      | 13            | 60            | 52  | 21  | 31   | 29  |
| Marital status (%)                  |                     | 68834   | 90   | 88   | 91            | 89                      | 88            | 90            | 88  | 91  | 88   | 91  |
| Highest education (%)               |                     | 17140   | 25   | 37   | 14            | 41                      | 45            | 9             | 23  | 27  | 30   | 32  |
| No formal school                    |                     | 17101   | 25   | 28   | 19            | 28                      | 31            | 17            | 16  | 35  | 24   | 28  |
| Primary school                      |                     | 26340   | 38   | 29   | 47            | 25                      | 21            | 53            | 39  | 35  | 35   | 32  |
| Middle and high school              |                     | 8252    | 12   | 6    | 20            | 6                       | 3             | 21            | 22  | 4   | 11   | 9   |
| College or university               |                     |         |      |      | 6             | 16                      | 28            | 8             | 10  | 22  | 21   | 15  |
| Annual household income (%)         |                     |         |      |      | 27            | 9                       | 26            | 9             | 28  | 8   | 10   | 22  |
| <¥12000                             | 11095               | 16      | 27   | 9    | 26            | 9                       | 28            | 8             | 10  | 22  | 21   | 15  |
| ¥12000-19999                        | 12475               | 18      | 21   | 15   | 23            | 16                      | 24            | 11            | 16  | 21  | 19   | 21  |
| ¥20000-59999                        | 25420               | 37      | 35   | 36   | 23            | 34                      | 36            | 35            | 35  | 40  | 35   | 39  |
| ¥60000-99999                        | 10148               | 15      | 10   | 19   | 9             | 19                      | 8             | 21            | 19  | 10  | 13   | 13  |
| ¥100000-199999                      | 7677                | 11      | 6    | 16   | 6             | 15                      | 4             | 19            | 16  | 6   | 10   | 9   |
| >¥200000                            | 1932                | 3       | 1    | 5    | 1             | 4                       | 1             | 6             | 5   | 1   | 2    | 3   |
| Occupation (%)                      |                     |         |      |      | 46            | 23                      | 50            | 24            | 54  | 15  | 18   | 56  |
| Primary industry                    | 24065               | 35      | 46   | 23   | 50            | 24                      | 54            | 15            | 18  | 56  | 39   | 38  |
| Secondary industry                  | 5464                | 8       | 9    | 7    | 8             | 5                       | 10            | 6             | 9   | 8   | 6    | 6   |
| Tertiary industry                   | 26388               | 38      | 30   | 46   | 28            | 45                      | 29            | 48            | 51  | 24  | 38   | 35  |
| Unemployed                          | 12860               | 19      | 15   | 24   | 15            | 23                      | 11            | 27            | 25  | 12  | 15   | 21  |

(continued on next page)
| Characteristic                            | No. of participants | Overall     | DASH Q1  | DASH Q5  | aMED Q1  | aMED Q5  | Sichuan Basin Q1  | Sichuan Basin Q5  | Yunnan-Guizhou Plateau Q1  | Yunnan-Guizhou Plateau Q5  | Qinghai-Tibet Plateau Q1  | Qinghai-Tibet Plateau Q5  |
|-----------------------------------------|---------------------|-------------|----------|----------|----------|----------|-------------------|-------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|
| Physical activity                       | 68556               | 23.7        | 26.7     | 21.0     | 25.8     | 22.2     | 27.2              | 17.8              | 18.8                         | 25.5                         | 22.2                        | 23.4                        |
| (MEIs-h/day) *                          | 68834               | 23.6        | 23.8     | 23.4     | 23.8     | 22.6     | 23.5              | 23.7              | 23.1                         | 23.7                         | 33.0                        | 23.5                        |
| BMI (kg/m²)                             |                     | (13.4, 38.7)| (13.9, 42.0)| (13.0, 34.6)| (13.5, 36.1)| (13.0, 33.6)| (13.0, 35.0)     | (13.0, 37.0)     | (18.2, 44.7)                 | (14.0, 39.9)                 | (18.2, 44.7)                | (12.1, 38.3)                |
| Total energy intake (kcal/day)          | 68834               | 1751.8      | 1665.8   | 1796.9   | 1691.1   | 1857.5   | 1407.7            | 1213.5            | 1569.2                       | 1427.5                       | 1809.9                     | 2225.1                      |
|                                         |                     | (215.1, 25.9)| (215.2, 26.2)| (211.4, 25.5)| (214.6, 26.2)| (215.6, 25.9)| (217.7, 25.9)    | (211.5, 25.6)     | (21.1, 25.6)                | (21.3, 25.6)                 | (1139.0, 1809.9)            | (128.3, 2703.6)             |
| Regular smoking (%)                     | 68832               | 15          | 9        | 23       | 8        | 21       | 10                | 19                | 3                            | 14                          | 13                          | 13                          |
| Dietar supplement (%)                   | 68832               | 15          | 9        | 23       | 8        | 21       | 10                | 19                | 3                            | 14                          | 13                          | 13                          |
| Regular sweeten beverage intake (%)     | 68834               | 63          | 94       | 94       | 87       | 89       | 89                | 87                | 95                          | 96                          | 82                          | 82                          |
| Current (kg/m²)                         | 68834               | 50         | 6       | 6       | 12       | 4        | 10                | 13                | 5                            | 17                          | 3                           | 17                          |
| Regular spicy food intake (%)           | 68834               | 79          | 78      | 81      | 71       | 85       | 71                | 70                | 84                          | 79                          | 73                          | 79                          |
| Regular pepper food intake (%)          | 68834               | 68          | 57      | 75      | 54       | 78       | 53                | 64                | 73                          | 58                          | 68                          | 68                          |
| Insomnia symptom (%)                    | 68772               | 42          | 47      | 38      | 45       | 40       | 45                | 40                | 42                          | 46                          | 40                          | 46                          |
| Depressive symptom (%)                  | 68774               | 4          | 7       | 3       | 6        | 3        | 6                 | 3                 | 5                           | 7                           | 3                           | 3                           |
| Anxiety symptom (%)                     | 68772               | 5          | 9       | 3       | 7        | 4        | 8                 | 3                 | 7                           | 8                           | 3                           | 3                           |
| Menopausal status in women (%)          |                     | 22439       | 52      | 44      | 58       | 45       | 56                | 43                | 61                          | 60                          | 47                          | 51                          |
| Premenopause                            | 3078                | 7          | 7       | 7       | 8        | 7        | 7                 | 7                 | 7                           | 7                           | 7                           | 7                           |
| Perimenopause                           | 17351               | 40         | 50      | 35      | 48       | 36       | 50                | 34                | 46                          | 42                          | 40                          | 40                          |
| Family history (%)                      |                     | 68834       | 32      | 25      | 39       | 21       | 39                | 36                | 27                          | 29                          | 29                          | 29                          |

Abbreviations: Q1 and Q5: the lowest and highest quintiles of dietary pattern scores; MEIs-h/day: hours of metabolic equivalent tasks per day; BMI: body mass index.

a. Data are presented as median (25th, 75th percentiles) for continuous variables and percentages (%) for categorical variables. We only display the results of Q1 and Q5 of dietary pattern scores for simplicity, with the corresponding median (25th, 75th percentiles) displayed in the following bracket.

b. Missing values are observed for marital status, highest education, annual household income, occupation, physical activity, dietary supplement, insomnia symptom, depressive symptom and anxiety symptom. The number of missing values per variable equal to the total number of participants (68834) minus the number presented in this column.

c. We aggregate various ethnic groups into three geographic regions due to their high similarity in dietary pattern and baseline characteristics.

d. Primary industry refers to occupations that involves getting raw materials, such as agriculture, forestry, fishing, and mining. Secondary industry refers to occupations that involves the transformation of the raw material into manufactured goods, such as factory worker. Tertiary industry refers to occupations that involves giving away direct services to its consumers, such as administrator, teacher, sales, etc.

e. Physical activity considers participants’ occupational, traffic, chores, and leisure time activities. We then calculated the hours of metabolic equivalent tasks per day (METs-h) for each participant.

f. Data are only available for women.

g. Family history refers to the self-reported hypertension, diabetes or cardiovascular disease from at least one first-degree relative (biological parents, sibling) in the baseline survey.
Figure 2. Estimated associations between various dietary patterns and cardiometabolic risks according to quintiles of dietary pattern scores, with the lowest quintile as reference group. Analyses were adjusted for sex, age, urbanicity, ethnicity, marital status, highest education attained, household income, profession, regular smoking, physical activity, total energy intake, regular intake of sweetened beverage, regular intake of dietary supplements, regular intake of spicy food, regular intake of pepper food, insomnia symptoms, depressive symptom, anxiety symptom, menopause status for women, and family history of cardiometabolic diseases using the inverse probability of exposure weighting. N in the brackets represent the number of newly diagnosed cardiometabolic risks. Q2-Q5 represent the second to fifth quintiles of dietary pattern scores. The filled blue dots represent adjusted odds ratios and the vertical blue lines represent 95% confidence intervals.

Our conclusions were robust with regard to all types of sensitivity analyses (Appendix text 14s). As expected, the inverse associations of DASH and positive associations of the Yunnan-Guizhou plateau dietary pattern with cardiometabolic risks (hypertension and MetS) were slightly attenuated but remained statistically significant when we further included self-reported cardiometabolic outcomes or adjusted for BMI as a confounder, which could be attributable to the potential inverse causality and mediation effects of BMI, respectively.

Discussion

Summary of main results

In this CMEC study covering diverse less-developed ethnic regions, greater adherence to DASH but not aMED was consistently associated with lower cardiometabolic risks, especially for hypertension and MetS. In contrast, two localized a posteriori dietary patterns in LEMRs were associated with increased cardiometabolic risks. Those associations were consistent in various subgroups de-
A. Hypertension

| Subgroup           | No. of Events | DASH Odds Ratio (95% CI) | Yunnan-Guizhou Plateau dietary pattern Odds Ratio (95% CI) |
|--------------------|---------------|--------------------------|----------------------------------------------------------|
| Sex                |               |                          |                                                          |
| Male               | 6273          | 0.72 (0.87; 0.80)        | 1.45 (1.39; 1.51)                                        |
| Female             | 6530          | 0.73 (0.87; 0.78)        | 1.41 (1.36; 1.45)                                        |
| Heterogeneity test |               | 0.90 (0.89; 0.91)        | 2.39 (2.36; 2.42)                                        |
| Age(years)         |               |                          |                                                          |
| <60                | 8326          | 0.69 (0.84; 0.74)        | 1.48 (1.42; 1.53)                                        |
| >60                | 4477          | 0.70 (0.86; 0.83)        | 1.49 (1.43; 1.53)                                        |
| Heterogeneity test |               | 1.21 (1.20; 1.22)        | 2.02 (1.99; 2.05)                                        |
| Regular smoking    |               |                          |                                                          |
| Never              | 8961          | 0.73 (0.86; 0.79)        | 1.37 (1.28; 1.46)                                        |
| Previous           | 756           | 0.76 (0.86; 0.84)        | 1.59 (1.52; 1.65)                                        |
| Current            | 3096          | 0.70 (0.86; 0.85)        | 1.44 (1.37; 1.51)                                        |
| Heterogeneity test |               | 0.90 (0.90; 0.90)        | 2.39 (2.36; 2.42)                                        |
| Physical activity (MET hours/day) |       |                          |                                                          |
| <23.70             | 6893          | 0.72 (0.86; 0.80)        | 1.47 (1.41; 1.53)                                        |
| ≥23.70             | 5064          | 0.71 (0.85; 0.77)        | 1.31 (1.25; 1.37)                                        |
| Heterogeneity test |               | 0.90 (0.90; 0.90)        | 2.39 (2.36; 2.42)                                        |
| Urbanicity         |               |                          |                                                          |
| Rural              | 8037          | 0.70 (0.83; 0.85)        | 1.24 (1.17; 1.31)                                        |
| Urban              | 4786          | 0.65 (0.78; 0.72)        | 1.50 (1.42; 1.57)                                        |
| Heterogeneity test |               | 0.90 (0.90; 0.90)        | 2.39 (2.36; 2.42)                                        |
| Ethnic group       |               |                          |                                                          |
| Sichuan Basin      | 6217          | 0.65 (0.80; 0.71)        | 1.59 (1.51; 1.67)                                        |
| Qianghai-Tibet Plateau | 980         | 0.90 (0.73; 1.12)        | 1.53 (1.46; 1.61)                                        |
| Yunnan-Guizhou Plateau | 5626      | 0.80 (0.73; 0.87)        | 1.20 (1.13; 1.27)                                        |
| Heterogeneity test |               | 0.90 (0.90; 0.90)        | 2.39 (2.36; 2.42)                                        |

B. MetS

| Subgroup           | No. of Events | DASH Odds Ratio (95% CI) | Yunnan-Guizhou Plateau dietary pattern Odds Ratio (95% CI) |
|--------------------|---------------|--------------------------|----------------------------------------------------------|
| Sex                |               |                          |                                                          |
| Male               | 3842          | 0.84 (0.76; 0.94)        | 1.24 (1.12; 1.38)                                        |
| Female             | 4556          | 0.76 (0.72; 0.87)        | 1.27 (1.16; 1.39)                                        |
| Heterogeneity test |               | 0.90 (0.89; 0.91)        | 2.39 (2.36; 2.42)                                        |
| Age(years)         |               |                          |                                                          |
| <60                | 6235          | 0.71 (0.79; 0.64)        | 1.32 (1.21; 1.43)                                        |
| >60                | 1963          | 0.82 (0.70; 0.95)        | 1.59 (1.47; 1.74)                                        |
| Heterogeneity test |               | 0.90 (0.56; 0.54)        | 2.39 (2.36; 2.42)                                        |
| Regular smoking    |               |                          |                                                          |
| Never              | 5863          | 0.80 (0.74; 0.86)        | 1.36 (1.25; 1.48)                                        |
| Previous           | 430           | 0.96 (0.89; 1.03)        | 1.52 (1.42; 1.63)                                        |
| Current            | 1965          | 0.85 (0.75; 0.95)        | 1.40 (1.31; 1.51)                                        |
| Heterogeneity test |               | 0.90 (0.46; 0.48)        | 2.39 (2.36; 2.42)                                        |
| Physical activity (MET hours/day) |       |                          |                                                          |
| <23.70             | 4957          | 0.76 (0.88; 0.84)        | 1.26 (1.14; 1.39)                                        |
| ≥23.70             | 3641          | 0.78 (0.81; 0.87)        | 1.01 (0.90; 1.13)                                        |
| Heterogeneity test |               | 0.90 (0.29; 0.26)        | 2.39 (2.36; 2.42)                                        |
| Urbanicity         |               |                          |                                                          |
| Rural              | 5200          | 0.81 (0.74; 0.89)        | 1.25 (1.14; 1.37)                                        |
| Urban              | 2988          | 0.76 (0.86; 0.84)        | 1.00 (0.89; 1.13)                                        |
| Heterogeneity test |               | 0.90 (0.29; 0.26)        | 2.39 (2.36; 2.42)                                        |
| Ethnic group       |               |                          |                                                          |
| Sichuan Basin      | 3948          | 0.73 (0.80; 0.81)        | 1.41 (1.32; 1.50)                                        |
| Qianghai-Tibet Plateau | 626         | 0.90 (0.75; 1.27)        | 1.51 (1.41; 1.62)                                        |
| Yunnan-Guizhou Plateau | 3924      | 0.62 (0.61; 0.76)        | 0.90 (0.81; 1.11)                                        |
| Heterogeneity test |               | 0.90 (0.29; 0.26)        | 2.39 (2.36; 2.42)                                        |

Figure 3. Stratified analysis of estimated associations between various dietary patterns and cardiometabolic risks according to predefined characteristics, by comparing the highest with the lowest quintiles. Analyses were adjusted for sex, age, urbanicity, ethnicity, marital status, highest education attained, household income, profession, regular smoking, physical activity, total energy intake, regular intake of sweetened beverage, regular intake of dietary supplements, regular intake of spicy food, regular intake of pepper food, insomnia symptoms, depressive symptom, anxiety symptom, menopause status for women, and family history of cardiometabolic diseases using the inverse probability of exposure weighting, with exclusion of the stratified variable as appropriate. Chi-square tests ($\chi^2$) were performed to examine heterogeneity among different subgroups. The filled blue dots represent adjusted odds ratios and the vertical blue lines represent 95% confidence intervals.

fined by sex, age, smoking and physical activity, but with the magnitudes that differed substantially across different ethnic regions and urbanicity. If those associations were mainly causal, DASH-like diets would be a superior dietary recommendation to reduce cardiometabolic risks in settings similar to the LEMRs in CMEC.

Substantial SES and ethnic disparities

In recent decades, diet quality has experienced global trends similar to those of cardiometabolic disease mortality.\(^5\) Accordingly, suboptimal diets have now become one of the major driving factors for the rise of cardiometabolic diseases in LMICs.\(^5\) Although the Chinese government has done remarkable achievements in improving the living standards in LEMRs, our findings still found the huge SES and ethnic disparity in diet quality and related cardiometabolic risks. Compared to the Han majority in the more developed Sichuan Basin, ethnic minorities in less developed plateau areas generally adhered to much worse dietary patterns, which contributed to a substantial increase in cardiometabolic risks. A recent systematic analysis of the global burden of disease reported that high sodium intake and low whole-grain and fruit intake were the leading three dietary risk factors for death and DALYs world-
wide. In China, national nutrition surveys have shown that high consumption of sodium and low consumption of fruits are associated with the greatest number of cardiometabolic deaths. Unfortunately, we can see the suboptimal intakes of sodium, whole grain and fruit simultaneously in LEMRs compared to the more developed region in our study (i.e. Sichuan basin). Given that China already has the highest rates of diet-related cardiometabolic disease death worldwide, there is an urgent need for dietary actions to prevent the ongoing epidemics of cardiometabolic diseases in LEMRs.

**Hypertension is the top priority**

It is well acknowledged that suboptimal diets are associated with a myriad of cardiometabolic risks. However, in this study, hypertension manifested a substantially stronger association with dietary patterns than the other three included cardiometabolic risks, suggesting that hypertension is the predominant diet-related cardiometabolic risk in LEMRs. Although MetS (a combined indicator of multiple cardiometabolic risks) was also considerably associated with dietary patterns, the majority of its association can be attributable to hypertension as well, given that only weak associations were observed between dietary patterns and other cardiometabolic risks, such as diabetes and dyslipidemia. These results were not surprising because hypertension in China have been a major concern for a long time. Our study indicates that hypertension control should be the top priority target among various diet-related cardiometabolic risks in LEMRs.

**Weight loss may be not an ideal control target**

In the development of cardiometabolic diseases, it is well recognized that obesity plays a crucial role and is highly associated with multiple cardiometabolic risks. Taking hypertension as an example, previous population studies suggest that overweight can directly account for 65% to 75% of primary hypertension. A growing number of studies also indicate that overweight may be an important mediator of the association between diet and cardiometabolic risks. In this CMEC study with a relatively lean population, we found that overweight only mediated 10% of the associations between dietary patterns and hypertension but that the overwhelming majority of the risk of hypertension was attributable to the direct effects of dietary patterns. Due to the high correlation between central obesity (prerequisite for MetS) and overweight, the proportions of mediations were expected to be relatively high for MetS, but the major part of the risk of MetS (approximately 65%) was still attributable to the direct effects of dietary patterns. Given that the total energy content of a diet is the main determinant of an overweight status, our results imply that the principal risk of diet on cardiometabolic health in the CEMC study may derive from dietary components directly rather than energy imbalance. Therefore, weight loss or solely energy limit may not be an ideal control target for diet-related cardiometabolic risks in LEMRs.

**DASH offers superior dietary guidance compared to aMED**

DASH and Mediterranean style diets are the most well-studied healthy dietary patterns for reducing cardiometabolic risks. In our study, DASH was consistently and inversely associated with various cardiometabolic risks, in line with multiple previous systematic reviews. It is also worth noting that dairy products contribute a predominate proportion to the beneficial effects of DASH. To date, evidence from both systematic reviews and Mendelian randomization studies in HICs have only suggested weak inverse associations between dairy products and cardiometabolic risk, but few of those studies were from LMICs, especially in those LMICs where dairy consumption is considerably low, such as China, Southeast Asia and Africa. Our findings indicate that dairy products may produce strong protective effects on cardiometabolic risks in LEMRs with low levels of dairy consumption, which coincides with the findings from the Prospective Urban Rural Epidemiology (PURE) study and another cohort study in China. According to previous systematic reviews, the beneficial effects of Mediterranean-style diets compared to DASH on cardiometabolic risks were controversial, especially for non-Mediterranean populations and racial/ethnic minority groups. In our study, we only observed weak inverse associations of aMED with hypertension and diabetes and even a positive association between aMED and MetS. As the major features of Mediterranean-style diets, the component of MUFA: SFA ratio did not show the expected beneficial effects on cardiometabolic risks based on single-component analyses. As another ethnic minority study suggested, the MUFA: SFA ratio may represent types of oils other than the intended healthy olive oil in non-Mediterranean populations. Similarly, we found that the MUFA: SFA ratio can only reflect a high consumption of vegetable oils (Appendix table 6s) due to a serious lack of high-quality sources of MUFA (such as olive oil or marine fish) in LEMRs. Besides, the stir-frying cooking manner with high temperature in LEMRs could be another reason why we did not observe the beneficial effects of the MUFA: SFA ratio in this study. In summary, our results indicate that DASH provides superior dietary guidance compared to aMED regarding reducing cardiometabolic risks in LEMRs.

**Remaining issues related to translating dietary guidance into practices**

Owing to the poor availability and affordability of cardiometabolic medicines in LMICs, dietary intervention is still
an effective and feasible way to prevent ongoing cardiometabolic epidemics.\textsuperscript{82,83} However, adherence to healthy dietary guidance heavily relies on access to high-quality food.\textsuperscript{84} For the highly geography-related dietary patterns in our study, the food choices in LEMRs were confined by the relatively infertile and inconvenient plateau environment. The crux of translating healthy dietary guidance into practices is still the availability and affordability of high-quality food. This may explain why stronger inverse associations of DASH with cardiometabolic risks were found among the Han majority in more developed regions than ethnic minorities in less developed regions. Deep-rooted culinary culture is another challenge to reduce the diet-related cardiometabolic risks in LEMRs. Although a global shift in dietary patterns has been seen worldwide,\textsuperscript{59} most ethnic minorities still live distinct lifestyles with unique culinary cultures that can hardly be altered. Thus, healthy dietary guidance is far from sufficient, and multi-sectional cooperation in refining food quality and acceptable dietary guidance that are suitable for local culture are remaining challenges for improving diet-related cardiometabolic health in LEMRs.

**Strengths and limitations**

To our knowledge, this is the first large-scale epidemiological study to comprehensively examine associations between dietary patterns and cardiometabolic risks in LEMRs. Our study provides a unique opportunity to understand the diet quality and related cardiometabolic risks in LEMRs, which has rarely been performed. We used standardized and validated methods to measure diet and objective indicators to record cardiometabolic outcomes. In addition, we performed analyses under the framework of causal inference, including the application of ESC-DAGs and marginal structure models, which provides transparent and evidence-based approaches to guide the selection of potential confounders and estimation methods. Nonetheless, limitations are worth noting. First, for the sake of feasibility, the FFQ used in our study only includes thirteen crude food groups rather than specific food items. Although it should not have a significant impact on the assessment of dietary patterns, crude food groups may undermine the accuracy of the calculation of particular dietary components (i.e., the MUFA: SFA ratio in aMED) and total energy intake. Our assessment tool is imperfect, but this simple tool might be the only feasible way to collect dietary information in LEMRs. For those participants from LEMRs, many of them are illiterate (they cannot read or fill the food questionnaire by themselves), speak different local languages (interviews need help from the local translator) and consume distinct foods (many foods are not included in any existing food database). Given the huge number of participants recruited and comprehensive other information collected, a food group-based and simplified food questionnaire may be the only option that can ensure the efficiency of communication, the cooperation of participants, and the comparability among various regions. By analysing subsample data for 24 HDRs, differences between the calculation of total energy intakes based on crude food groups and that based on specific food items were small and symmetrically distributed around zero (Appendix text 15s). Second, we modified the dietary pattern scores by omitting SSBs from the DASH score and nuts from the MED score, which may attenuate the associations between dietary scores and cardiometabolic risks. For omitting the consumption of nuts, it may also lead to the underestimation of MUFA: SFA ratio. Nevertheless, we believe that this defect should not have a marked impact on our conclusion given that the consumption of both SSBs and healthy nuts in LEMRs was very low. Third, we used questions on household use as a proxy for individual-level intake to estimation both sodium and oil intake. In China, people always consumed mixed dishes and shared dishes with each other, and the standard recipes with amounts of condiments specified precisely is also rarely used, which make it infeasible to recall and estimate salt and oil intake by questionnaire. Fourth, we did not perform measurement error correction, as the overwhelming majority of current statistical approaches are only suitable for a single food item or nutrient.\textsuperscript{85,86} By conducting a preliminary simulation study, we found that the dietary patterns were robust to the measurement errors and that the impact of measurement errors on the diet-disease association was still the attenuation effect (Appendix text 16s). Fifth, dietary pattern analysis can capture only a portion of the variation in food consumption\textsuperscript{87} and therefore may not fully represent diet quality. Sixth, residual confounding is still possible, even though we carefully adjusted for potential confounders under the framework of causal inference. Last, the intrinsic nature of the cross-sectional study design also limits the reliable inference of causality, but we did exclude self-reported cases to eliminate potential reverse causality.

**Conclusion**

In conclusion, using the recently collected CMEC data, we observed substantial SES and ethnic disparities in diet quality and related cardiometabolic risks in LEMRs. Among the four included cardiometabolic risks, hypertension showed the strongest association with dietary patterns and should be the top priority of control target for ensuing dietary actions. Our study also suggests that DASH offers superior dietary guidance to aMED for reducing cardiometabolic risks in LEMRs. In particular, the dairy intake encouraged by DASH may produce considerable beneficial effects for lowering cardiometabolic risks. Although the spectrum of cardiometabolic diseases and dietary patterns may vary among different LEMRs, our study reaffirms the important role of diet in the prevention of cardiometabolic diseases and DASH may provide a solution of dietary guidance with availability and affordability in the setting of LEMRs.

**Contributors**

XX, JY, and XZ contributed to the design of the present study. XZ was the principal investigator and JY was the co-principal investigator of the CMEC study. XX and ZQ wrote the analysis plan, and the first and final draft of the paper. JY and XZ reviewed and commented on the data analysis, all drafts and the final paper. All other authors were involved in conduct of the study, analysis of data, interpretation of results, and provided critical comments on all drafts of the report.

**Research in context**

**Evidence before this study**

We searched for PubMed for relevant literature between Jan 1, 1980, and Feb 1, 2021, using the search terms “dietary pattern” or “eating pattern” or “dietary index” or “dietary guidance” or “diet quality” or “dietary approaches to stop hypertension” or “Mediterranean” and “cardiovascular” or “metabolic” or “cardiometabolic” or “heart” or “atherosclerosis” or “stroke” or “hypertension” or “diabetes” or “hyperlipidaemia”. Few studies on the associations between dietary patterns and cardiometabolic risks have been conducted in less-developed ethnic minority regions (LEMRs).

**Added value of this study**

To our knowledge, this is the first large-scale epidemiological study to comprehensively examine associations between dietary patterns and cardiometabolic risks in LEMRs. Our study provides
a unique opportunity to understand diet quality and related cardiometabolic risks in LEMRs, which has rarely been conducted previously.

Implications of all the available evidence

Given the substantial socioeconomic status and ethnic disparities in diet quality and related cardiometabolic risks shown in this study, there is an urgent need for dietary actions to prevent the ongoing epidemics of cardiometabolic diseases in LEMRs, especially hypertension. Our findings also suggest that Dietary Approaches to Stop Hypertension (DASH) offers superior dietary guidance compared to alternative Mediterranean (aMED)-style diets for reducing cardiometabolic risks in LEMRs.

Declaration of interests

All authors declared no competing interest.

Declaration of Competing Interest

The authors declare no conflict of interest.

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Data sharing

Study data are available on request to the authors.

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Supplementary materials

Supplementary material associated with this article can be found, in the online version, at doi: 10.1016/j.jamipal.2021.100252.

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