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Association between branched-chain amino acid levels and gastric cancer risk: large-scale prospective cohort study

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Background: Gastric cancer (GC) remains a malignancy with high incidence and mortality rates worldwide. Although branched-chain amino acids (BCAAs) play a crucial role in various physiological and pathological processes, their specific relationship with risk of GC remains unclear.

Methods: We conducted a large-scale prospective cohort from UK Biobank database. We evaluated the relationship between BCAA levels and risk of GC using Cox regression, Kaplan–Meier survival curves, the accelerated failure time (AFT) model, and restricted cubic spline (RCS) analysis.

Results: During the follow-up of 12 years, 247,753 participants were included in the study. And the Cox regression analysis revealed that higher levels of isoleucine (HR = 0.65, 95% CI 0.48–0.89; p = 0.007), leucine (HR = 0.57, 95% CI 0.42–0.79; p < 0.001), valine (HR = 0.53, 95% CI 0.39–0.73; p < 0.001), and total BCAAs were associated with a reduced risk of GC (HR = 0.51, 95% CI 0.37–0.70; p < 0.001). Kaplan–Meier curves and the AFT model confirmed that elevated BCAA levels significantly delayed the onset of GC. Additionally, RCS analysis identified nonlinear dose–response relationships between BCAAs and risk of GC. Stratified analyses indicated that the protective effect of BCAAs was consistent across various subgroups, with a more pronounced impact in older individuals without chronic diseases.

Conclusion: Elevated BCAA levels are significantly associated with a reduced risk of GC, particularly in older adults. This finding highlights the potential of BCAAs in GC prevention and suggests that future research and clinical practice should emphasize regulating BCAA levels.

KEYWORDS

amino acid, gastric cancer, UK Biobank, prospective cohort study, cancer prevention

Introduction

Gastric cancer (GC) is a prevalent and highly lethal malignancy worldwide. According to the Global Cancer Observatory (GLOBOCAN), >1 million new cases of GC were reported globally in 2022, accounting for 5.6% of all new cancer cases. The mortality rate for GC remains high, with approximately 770,000 deaths annually, making it the third leading cause of cancer-related deaths (1, 2). Many patients are diagnosed at an advanced stage due to the asymptomatic nature of the early disease, leading to poor treatment outcomes and prognosis (3). Despite advancements in early screening and therapeutic approaches, the overall survival rate for GC remains low (4). Therefore, investigating the pathogenesis and risk factors of GC is crucial for identifying effective prevention and treatment strategies.

Amino acids, as fundamental building blocks of proteins, are crucial in sustaining life and regulating bodily functions (5-7). Recent studies have revealed that amino acids are vital not only for metabolic regulation, cellular signaling, and immune responses but also in the onset and progression of various cancers (8, 9). For example, certain amino acids can promote cancer cell growth and proliferation, influence the tumor microenvironment, and serve as potential biomarkers for early cancer detection and prognosis assessment (10, 11). Branched-chain amino acids (BCAAs), including leucine, isoleucine, and valine, are essential amino acids that have garnered increasing attention in research on metabolic diseases and cancer. Despite their recognized roles in other types of cancers, the specific functions and underlying mechanisms of BCCAs in GC remain unclear. Preliminary studies suggest that BCAAs may be involved in the metabolic regulation and proliferation of GC cells, but the detailed mechanisms and effects require further exploration (12). Therefore, a systematic investigation of the relationship between BCAAs and GC is crucial not only to address this research gap but also to potentially uncover new insights and approaches for the prevention and treatment of GC.

Therefore, exploring the correlation between BCAAs and the risk of developing GC holds significant scientific and clinical value. This study, using a large-scale prospective cohort, aims to systematically evaluate the impact of BCAAs on GC risk. The findings will provide a theoretical basis and new research directions for future strategies in the prevention and treatment of GC.

Methods

Study population

A total of 502,357 volunteers, aged 37–73 years, were enrolled in UK Biobank (UKB) between 2006 and 2010. Data were collected through a self-administered touchscreen questionnaire, a brief interview, physical and functional measurements, and biological samples collected during the assessment visit. Detailed information has been reported elsewhere (13).

Measurement of amino acids

Amino acids data were obtained from a nuclear magnetic resonance (NMR) platform, which measured 251 metabolic indicators

in >250,000 individuals across two phases: phase 1 (June 2019 to April 2020) and phase 2 (April 2020 to June 2022) (14, 15). This platform provides measurements for nine amino acids, including BCAA. Since the study is based on the data available from the platform, we included all amino acid markers provided to maximize the use of available data for a comprehensive analysis of GC risk. The BCAAs included isoleucine, leucine, and valine. The total BCAA measurement was calculated as the sum of these three amino acids.

Assessment of GC

Diagnosis of GC was determined using International Classification of Diseases, Tenth Revision (ICD-10) codes C16.0–16.9 from the cancer registry. Data on death cases were obtained from death registries. The follow-up period for participants began on their enrollment date and continued until the earliest of the following events: the first occurrence of GC or another cancer, the participant's death, or the study cutoff date (June 1, 2022).

Assessment of other covariates

Researchers collected participants' demographic and medical information through interviews and touchscreen questionnaires during the baseline visit. This information included age, sex, ethnicity, body mass index (BMI), physical activity levels, food intake frequency, Townsend Deprivation Index (TDI), smoking and drinking habits, history of cardiovascular disease (CVD) and diabetes mellitus (DM), and family history of cancer. Information on the use of lipid-lowering drugs was also recorded. Physical activity was measured in the metabolic equivalent task (MET) minutes (16). The TDI was used to represent socioeconomic status (17). The food intake frequency was used to calculate a diet score ranging from 0 to 9 points, with higher scores indicating poorer dietary habits (18). A family history of cancer was defined as having either parent with a cancer diagnosis. Missing values for baseline covariates were addressed using multiple imputations.

Selection criteria

Participants with incomplete recruit information and a cancer history at baseline were excluded (n=45,779). Additionally, participants with missing values for any amino acids were excluded (n=208,826). In total, 247,753 participants were included in this study.

Statistical analysis

BCAA levels were categorized into quartiles (Q1–Q4) to evenly divide the data sample, facilitating a more intuitive observation of how the risk of GC changes with increasing BCAA levels. Specifically, the quartiles were automatically calculated based on the distribution of BCAA levels in the dataset, with Q1 representing the lowest 25% of individuals in terms of BCAA levels, and Q4 representing the highest 25%. This grouping ensures that the number of individuals in each group is approximately equal, thereby reducing potential bias caused

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by differences in sample size between groups. Descriptive statistics were used to summarize and compare the baseline characteristics of participants across different total BCAA quartiles. Continuous variables were expressed as medians and interquartile ranges, while categorical variables were presented as frequencies and proportions. Group comparisons were performed using ANOVA for continuous variables and the chi-squared test for categorical variables.

To assess the relationship between amino acids and GC risk, multivariable Cox proportional hazard models were used to calculate hazard ratios (HRs) and 95% confidence intervals (CIs). The proportional hazards assumption was verified using Schoenfeld residuals. Amino acid data were categorized into quartiles to examine the effect of quartile increases on GC risk and to test for trends in association. The data were then transformed into z-scores, and HRs were calculated for each 1-unit increase in standard deviation (SD). Two models were established: model 1 adjusted for age, sex, and ethnicity, and model 2 further adjusted for diet score, TDI, MET, BMI, smoking and drinking status, DM, hypertension, CVD, lipid use, and family history of cancer. The cumulative Hazard Kaplan–Meier (KM) Curve was created to assess the hazard of GC across different amino acids quartiles.

Restricted cubic splines (RCS) were used to examine the dose– response relationship between amino acid levels and GC events, with knots placed at the 10th, 50th, and 90th percentiles of amino acid concentrations (19). Subsequently, an accelerated failure time (AFT) model was used to assess the impact of amino acid levels on the timing of GC events. The AFT model, using a Weibull distribution and the lowest quartile (Q1) of amino acids as the reference, evaluated the effects of increased amino acid levels on the acceleration or delay of GC occurrence (20).

To assess potential differences in study outcomes across various population groups, we stratified by age, sex, race, BMI, hypertension, DM, CVD, diet score, family history of cancer, and smoking and alcohol status. We then repeated the primary analyses. Several sensitivity analyses were also conducted to evaluate the robustness of the study findings. First, participants with a follow-up period of <2 years were excluded to mitigate the effect of reverse causality. Second, missing baseline covariate data were removed to assess the impact of imputation on the findings. Third, to account for death as a competing event, the Fine-Gray competing risks model and its sub-distribution were used to describe the impact of amino acid levels on GC risk in the context of competing death states (21). Finally, results from four additional datasets created through multiple imputations were analyzed and summarized to confirm the consistency of the study findings. Statistical analyses were performed using R (v. 4.3.1; R Foundation for Statistical Computing), with statistical significance defined as a two-sided p < 0.05.

Results

Baseline characteristics

Participants were categorized into quartiles based on their total BCAA levels, revealing significant statistical differences across

Characteristic	Quintile 1	Quintile 2	Quintile 3	Quintile 4	<i>p</i> -Value	
	(≤ 0.306)	(0.306–0.354)	(0.354–0.412)	(≥ 0.412)		
Age, years	57.0 (49.0-63.0)	58.0 (50.0-63.0)	58.0 (50.0-63.0)	57.0 (50.0-63.0)	< 0.001	
Male, N (%)	16,393 (26.5%)	26,947 (43.5%)	34,677 (56.0%)	38,648 (62.4%)	< 0.001	
White, <i>N</i> (%)	59,156 (95.5%)	58,828 (95.0%)	58,514 (94.5%)	58,286 (94.1%)	< 0.001	
BMI	25.1 (22.8–27.9)	26.4 (24.0-29.4)	27.4 (24.9-30.4)	28.2 (25.6–31.4)	< 0.001	
MET	1893.0 (885.0-3772.0)	1826.0 (836.0-3652.0)	1765.0 (792.0-3519.0)	1693.0 (744.0-3420.0)	< 0.001	
Townsend deprivation index	-2.1 (-3.6-0.5)	-2.2 (-3.7-0.5)	-2.2 (-3.7-0.4)	-2.2 (-3.7-0.5)	< 0.001	
Diet score	5.0 (4.0-6.0)	5.0 (4.0-6.0)	5.0 (4.0-6.0)	5.0 (4.0-6.0)	< 0.001	
DM	1,463 (2.4%)	2,181 (3.5%)	3,405 (5.5%)	5,853 (9.4%)	<0.001	
Hypertension	13,602 (22.0%)	16,241 (26.2%)	18,453 (29.8%)	20,330 (32.8%)	< 0.001	
CVD	3,804 (6.1%)	4,512 (7.3%)	5,198 (8.4%)	6,045 (9.8%)		
History of cancer family	18,441 (29.8%)	18,448 (29.8%)	18,550 (29.9%)	18,499 (29.9%)	0.898	
Lipid-lowering drugs, N (%)	3,217 (5.2%)	5,490 (8.9%)	7,855 (12.7%)	9,999 (16.1%)	<0.001	
Drinking status, N (%)						
Never	2,741 (4.4%)	2,702 (4.4%)	2,669 (4.3%)	2,697 (4.4%)		
Previous	2,298 (3.7%)	2052 (3.3%)	2082 (3.4%)	2,311 (3.7%)		
Current	56,897 (91.9%)	57,186 (92.3%)	57,187 (92.3%)	56,930 (91.9%)		
Smoking status, N (%)						
Never	26,052 (42.1%)	25,139 (40.6%)	24,666 (39.8%)	24,215 (39.1%)		
Previous	28,760 (46.4%)	30,269 (48.9%)	30,971 (50.0%)	31,618 (51.0%)		
Current	7,124 (11.5%)	6,532 (10.5%)	6,301 (10.2%)	6,105 (9.9%)		

TABLE 1 Stratified baseline characteristics according to total BCAA quintile levels.

BCAA, Branched-chain amino acid; BMI, body mass index; MET, metabolic equivalent task; DM, diabetes mellitus; CVD, cardiovascular diseases.

groups (p < 0.05). As BCAA levels increased, the proportion of males and BMI levels increased significantly, while MET levels decreased gradually. Furthermore, the prevalence of diabetes, hypertension, and CVD increased with increasing BCAA levels. Although dietary scores remained consistent across the groups, the proportion of participants using lipid-lowering medications significantly increased with increasing BCAA levels. In terms of smoking and drinking status, the proportion of current smokers slightly decreased as BCAA levels increased, whereas the proportion of former smokers correspondingly increased. These baseline characteristics suggest a significant association between BCAA levels and various demographic and clinical variables, providing crucial background information for further investigation into the relationship between BCAA levels and the risk of developing GC (Table 1).

Association between BCAAs and GC risk

Cox regression analysis revealed a significant inverse association between higher levels of BCAAs and the risk of GC. In contrast, no such relationship was observed for other amino acids (alanine, glutamine, glycine, histidine, phenylalanine, and tyrosine) (Supplementary Table S1). After adjusting for multiple confounding factors (Model 2), participants in the highest quartile (Q4) of isoleucine, leucine, and valine levels exhibited a 35% (HR 0.65, 95% CI 0.48–0.89; *p*=0.007), 43% (HR 0.57, 95% CI 0.42–0.79; *p*<0.001), and 47% (HR 0.53, 95% CI 0.39–0.73; *p*<0.001) reduction in GC risk, respectively, compared to those in the lowest quartile (Q1). Furthermore, total BCAA levels were also associated with a reduced GC risk (HR 0.51, 95% CI 0.37–0.70; *p* < 0.001) (Table 2). Further analysis revealed that each SD increase in isoleucine, leucine, and valine levels corresponded to a 14% (HR 0.86, 95% CI 0.76–0.97; *p* = 0.012), 15% (HR 0.85, 95% CI 0.75–0.96; *p*=0.007), and 20% (HR 0.80, 95% CI 0.71–0.91; *p*<0.001) reduction in GC risk, respectively. Additionally, each SD increase in total BCAA levels was associated with an 18% reduction in GC risk (HR 0.82, 95% CI 0.72–0.93; *p* = 0.002) (Table 2). These results suggest that higher BCAA levels may serve as a significant protective factor against the risk of GC.

Association between BCAAs and cumulative GC risk

To further validate the association between BCAA levels and GC risk, KM survival curves were used to compare the cumulative risk across different BCAA quintiles. As shown in Figure 1, the cumulative risk of GC in the highest BCAA level group (Q4) for leucine, valine, and total BCAAs was significantly lower than in the lowest BCAA level group (Q1), with statistically significant differences (leucine p < 0.001, valine p < 0.001, and total BCAA p < 0.001). Moreover, the cumulative risk for the Q4 group remained consistently lower than that for the Q1 group over the follow-up period (Figure 1). Surprisingly, the lowest cumulative risk for isoleucine was observed in the Q3 group (p = 0.007). These findings further support the role of BCAAs as potential protective factors in

TABLE 2 The association between BCAAs and the risk of gastric cancer.

	Model 1		Model 2					
Туре	HR (95%CI)	p	HR (95%CI)	p				
Isoleucine								
Q1	Reference		Reference					
Q2	0.82 (0.61-1.1)	0.185	0.79 (0.59–1.07)	0.129				
Q3	0.66 (0.49-0.9)	0.009	0.63 (0.46-0.86)	0.003				
Q4	0.71 (0.52–0.96)	0.025	0.65 (0.48-0.89)	0.007				
<i>p</i> for trend	0.013		0.003					
Per SD increase	0.88 (0.78-0.99)	0.035	0.86 (0.76-0.97)	0.012				
Leucine								
Q1	Reference		Reference					
Q2	0.7 (0.52-0.96)	0.027	0.69 (0.51-0.95)	0.021				
Q3	0.77 (0.57-1.03)	0.081	0.74 (0.55-1)	0.052				
Q4	0.61 (0.44-0.83)	0.002	0.57 (0.42-0.79)	< 0.001				
<i>p</i> for trend	0.007		0.002					
Per SD increase	0.87 (0.77–0.97)	0.017	0.85 (0.75-0.96)	0.007				
Valine								
Q1	Reference		Reference					
Q2	0.67 (0.5-0.91)	0.01	0.65 (0.48-0.89)	0.007				
Q3	0.68 (0.51-0.92)	0.012	0.65 (0.48-0.88)	0.005				
Q4	0.59 (0.43-0.8)	< 0.001	0.53 (0.39-0.73)	< 0.001				
<i>p</i> for trend	0.002		< 0.001					
Per SD increase	0.84 (0.74–0.94)	0.003	0.8 (0.71-0.91)	< 0.001				
Total BCAA								
Q1	Reference		Reference					
Q2	0.63 (0.47-0.86)	0.003	0.62 (0.45-0.83)	0.002				
Q3	0.63 (0.47-0.85)	0.003	0.6 (0.44-0.81)	< 0.001				
Q4	0.56 (0.41-0.76)	< 0.001	0.51 (0.37-0.7)	< 0.001				
<i>p</i> for trend	< 0.001		< 0.001					
Per SD increase	0.85 (0.75-0.95)	0.007	0.82 (0.72-0.93)	0.002				

BCAA, Branched-chain amino acid; GC, gastric cancer; HR, hazard ratio; CI, confidence interval; SD, standard deviation; Model 1 adjusted for age, sex, and ethnicity; Model 2 further adjusted for diet score, TDI, MET, BMI, smoking and drinking status, DM, Hypertension, CVD, lipid, and family history of cancer.

reducing GC risk, demonstrating a clear trend where higher BCAA levels are associated with lower GC risk.

Dose–response relationships between BCAAs and GC

The RCS analysis revealed a significant nonlinear association between BCAA levels and GC risk (leucine *p* for nonlinearity=0.011, valine *p* for nonlinearity=0.026, and total BCAA *p* for nonlinearity=0.008). At lower BCAA levels, the risk of GC was higher. As BCAA levels increased, the risk of GC gradually decreased, with a marked reduction in GC incidence at higher BCAA levels (Figure 2). Interestingly, for isoleucine, there was an observed increase in GC risk



at excessively high levels (isoleucine *p* for non-linearity=0.016) (Figure 2). This nonlinear dose–response relationship highlights the potential protective role of BCAAs in GC prevention and suggests that BCAAs may offer optimal protective effects within a specific range. These findings provide valuable insights for exploring the mechanisms and determining the optimal dosages of BCAAs for GC prevention and treatment.

AFT analysis of BCAA levels and GC risk

The AFT model demonstrated that higher levels of BCAAs significantly delayed the onset of GC. However, it was observed that the average time to GC onset for participants in the highest quartile (Q4) of isoleucine occurred earlier compared to those in the third quartile (Q3) (Figure 3). Furthermore, compared to the lowest quartile (Q1), the median time differences for isoleucine in the Q2, Q3, and Q4 quartiles were 2.75 months, 21.49 months, and 68.71 months, respectively (p=0.036). For leucine, valine, and total BCAA, the median time differences in the Q4 quartile compared to Q1 were 88.28 months, 112.27 months, and 130.89 months, respectively (leucine p=0.003, valine p=0.002, and total BCAA p<0.001) (Supplementary Table S2). These results underscore the protective role of higher BCAA levels in delaying the onset of GC.

Sensitivity analysis

To validate the robustness of the Cox regression results, we conducted three sensitivity analyses and a Fine-Gray competing risk model analysis. After excluding participants who developed GC within 2 years, higher levels of isoleucine (HR 0.65, 95% CI 0.47-0.91; *p* = 0.012), leucine (HR 0.58, 95% CI 0.41–0.81; *p* = 0.002), valine (HR 0.51, 95% CI 0.36–0.72; p < 0.001), and total BCAAs (HR 0.51, 95% CI 0.36-0.72; p < 0.001) remained significantly associated with a reduced risk of GC (Supplementary Table S3). Trend analyses for all BCAAs showed significant *p*-values (isoleucine p = 0.008, leucine p = 0.006, valine p < 0.001, and total BCAA p < 0.001) (Supplementary Table S3). Furthermore, analyses excluding participants with missing baseline covariates and those using multiple imputation techniques consistently supported these conclusions, indicating that higher BCAA levels still significantly reduced GC risk. The risk ratios per SD increase further emphasized the significant protective effect of BCAAs (Supplementary Tables S4, S5). Moreover, the Fine-Gray competing risk model indicated that compared to the lowest quartile (Q1), the highest quartile (Q4) of isoleucine, leucine, valine, and total BCAA levels were associated with reductions in GC risk of 34, 41, 46, and 48%, respectively (Supplementary Table S6). These results are consistent with the Cox regression analysis and reinforce the reliability of the study's conclusions regarding the impact of BCCAs on GC risk.



Subgroup analysis of BCAA levels and GC risk

To further investigate the protective effects of BCAA levels across different demographic and clinical subgroups, we conducted stratified analyses. The results revealed that isoleucine, leucine, valine, and total BCAA levels did not exhibit significant interactions with GC risk across various subgroups based on gender, BMI, smoking status, drinking status, or the presence of diabetes and hypertension (Figure 4). However, significant heterogeneity was observed when stratifying by age. Higher BCAA levels were associated with a reduced risk of GC in participants older than 60 years, while the risk slightly increased in those younger than 60 years. Additionally, an interaction was observed for leucine concerning smoking status. Specifically, current smokers showed a 17% increased risk of GC, whereas the risk was significantly reduced among never smokers and former smokers (Figure 4).

Discussion

This study provides a comprehensive analysis of the effects of BCAAs on the risk of GC. The findings demonstrate that higher levels

of isoleucine, leucine, valine, and total BCAAs are significantly associated with a reduced risk of GC. Stratified analyses further reveal that age influences the relationship between BCAAs and GC risk, with the protective effects of BCAAs being particularly pronounced in older individuals. Additionally, elevated BCAA levels were found to significantly delay the onset of GC. These results offer novel scientific insights into GC prevention strategies and highlight the potential role of BCAAs in oncology. They suggest that modulation of BCAA levels could be a valuable approach for preventing and delaying the onset of GC.

Previous research has reported similar protective effects of BCAAs across various cancer types. For example, some studies have reported an association between high levels of BCAAs and a reduced risk of liver cancer (22). Moreover, BCAAs may help decrease the incidence of obesity-related cancers by modulating metabolic pathways and improving insulin sensitivity (23, 24). Research by Viana et al. (25) suggested that higher levels of leucine could shift tumor metabolism from glycolysis to oxidative phosphorylation components, leading to increased mRNA and protein expression of oxidative phosphorylation components and consequently reduced tumor invasiveness and metastasis. These findings are consistent with our results and support the potential protective role of BCAAs in lowering cancer risk. However, an alternative perspective is provided by Tang et al. (26), who reported that



leucine and arginine might promote colorectal cancer cell proliferation by activating the PI3K/Akt/Wnt/ β -catenin pathway. This discrepancy may be attributed to the use of microcystins (MCs) in their study. MCs are potent carcinogens produced by cyanobacteria and are known for their strong carcinogenic properties (27-29). Although the PI3K/Akt/ Wnt/β-catenin pathway is crucial in cell proliferation and cancer development, its activation and regulation are highly complex and influenced by various internal and external factors. Therefore, we propose that the complex role of BCAAs in different cancer types may be influenced by factors such as cancer type, tumor microenvironment, and the metabolic state of the patient. Several physiological and molecular pathways could be involved in the mechanisms by which BCAAs regulate GC risk. First, BCAAs, particularly leucine, modulate protein synthesis and cell growth through the activation of the mTOR (mammalian target of rapamycin) signaling pathway (30). mTOR is a crucial regulator involved in tumor cell growth, proliferation, and survival (31). While overactive mTOR signaling is associated with many cancer types, moderate mTOR activity may help maintain normal cell functions and prevent excessive proliferation of cancer cells (32). Second, BCAAs play a significant role in regulating energy metabolism. By enhancing insulin sensitivity and improving glucose metabolism, BCAAs help prevent obesity and diabetes, which are major risk factors for GC (24, 33). Additionally, BCAAs have antiinflammatory properties that can reduce chronic inflammatory responses in the body (34). Chronic inflammation is a key factor in cancer development and progression, thus reducing inflammation through BCAAs may indirectly lower cancer risk. Furthermore, BCAAs may also influence the regulation of the gut microbiota. Dysbiosis, or imbalance in the gut microbiota, is closely linked with various diseases, including cancer (35). Studies have shown that BCAAs can modulate gut microbiota composition, enhance gut barrier function, and reduce the proliferation of harmful bacteria, potentially contributing to a decreased risk of GC (36). Furthermore, it is noteworthy that the mechanisms through which BCAAs exert their effects may extend beyond direct influences on cell growth and metabolism to include the modulation of the host's immune system. Research indicates that BCAAs may play a role in antitumor immune responses by affecting immune cell function. For instance, BCAAs can influence the metabolic pathways of T cells, thereby promoting their function and proliferation and enhancing the body's immune surveillance against tumors (37). Additionally, BCAAs may regulate the activity of immunosuppressive cells, such as modulating the function of regulatory T cells (Tregs) and reducing their inhibitory effects on antitumor responses (38). These mechanisms may further elucidate the protective effects of BCAAs in reducing cancer risk.

In clinical practice, the supplementation or modulation of BCAAs could have considerable practical implications. Given the metabolic characteristics of BCAAs and their potential protective role in GC, developing personalized BCAA intervention strategies is crucial. These strategies should be tailored based on the patient's disease stage, metabolic status, and genetic background. For instance, in certain



The association between BCAAs and the risk of GC across different demographic and clinical subgroups. (A) Isoleucine. (B) Leucine. (C) Valine. (D) Total BCAA. Models were adjusted with age, sex, and ethnicity, BMI (body mass index), MET (metabolic equivalent task), TDI (Townsend Deprivation Index), smoking and drinking status, DM (diabetes mellitus) Hypertension, CVD, lipid, and family history of cancer. BCAA: Branched-chain amino acid.

patients with GC, especially those with abnormal BCAA levels or disrupted metabolism, adjusting BCAA levels might be necessary to support and enhance treatment efficacy. Conversely, in individuals without cancer or those at high risk, increasing BCAA intake may help reduce the risk of GC, thereby serving a preventive role. To achieve these potential clinical applications, further clinical trials are essential to validate the safety and efficacy of BCAA interventions and to determine the optimal intervention protocols and dosages. These studies should account for individual differences, including genomic information, metabolic state, and lifestyle factors, to ensure that interventions are tailored to the specific needs of diverse populations.

This study's strengths include its large-scale prospective design and long-term follow-up data, which enhance the statistical significance and robustness of the results. Accurate NMR

measurements of BCAA levels help mitigate recall bias from dietary questionnaires, thus strengthening the reliability of the conclusions. However, despite providing strong evidence regarding the relationship between BCAAs and GC risk, several limitations must be acknowledged. First, as an observational study, it cannot establish causality. Second, BCAA levels were measured only at baseline, precluding the assessment of their effects over time and how fluctuations in these levels might influence cancer risk. Third, the study population primarily consists of individuals from a specific region in the UK, which may limit the generalizability of the findings to other populations, particularly those from different ethnic or geographic backgrounds. Finally, despite adjusting for multiple confounders, some unmeasured variables, such as genetic background, gut microbiota composition, or specific dietary patterns, may still influence the results. Future research should employ longitudinal designs, include more diverse sample populations, and explore the specific mechanisms through which BCAAs affect different groups to better understand their role in cancer prevention.

In conclusion, this study found that higher levels of BCAAs are significantly associated with a reduced risk of GC, with particularly pronounced protective effects in older individuals and those without chronic diseases. These findings provide preliminary evidence for the potential use of BCAAs in the prevention and treatment of GC. However, further research is needed to elucidate the specific mechanisms and to determine the optimal methods for their use. Such research will facilitate the development of more personalized and effective intervention strategies for patients with cancer and high-risk populations, thereby advancing the field of GC prevention and treatment.

Data availability statement

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

Ethics statement

This study was obtained from the North West Multi-Center Research Ethics Committee (approval number: 11/NW/0382), and all participants provided written informed consent.

Author contributions

LY: Writing - original draft, Methodology, Formal analysis, Conceptualization. SB: Writing - original draft, Methodology,

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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Supplementary material

The Supplementary material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fnut.2024.1479800/ full#supplementary-material

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