Dietary Patterns Associated with Food Insecurity Predict a Worse Prognosis for U.S. Cancer Survivors: NHANES 1999-2018

Christian A. Maino Vieytes1, Ruoqing Zhu2, Francesca Gany3, Brenda D. Koester4, Anna E. Arthur5

1 Division of Nutritional Sciences, The University of Illinois at Urbana-Champaign, Urbana, IL 61801, USA; [cam17@illinois.edu](mailto:cam17@illinois.edu)

2 Department of Statistics, The University of Illinois at Urbana-Champaign, Urbana, IL 61801, USA; [rqzhu@illinois.edu](mailto:rqzhu@illinois.edu)

3 Memorial Sloan Kettering Cancer Center, New York, NY 10065, USA; [ganyf@mskcc.org](mailto:ganyf@mskcc.org)

4 The Family Resiliency Center, The University of Illinois at Urbana-Champaign, Urbana, IL 61801, USA; [bkoester@illinois.edu](mailto:bkoester@illinois.edu)

5 Department of Dietetics and Nutrition, The University of Kansas Medical Center,   
Kansas City, KS 66160, USA; [aarthur4@kumc.edu](mailto:aarthur4@kumc.edu)

**Abstract**

Background

Food insecurity—the lack of unabated access to nutritious foods—is a consequence many cancer survivors face. Food insecurity is associated with adverse health outcomes and lower diet quality among the general public. In a previous analysis, we extracted prevailing dietary patterns in U.S. cancer survivors suggesting poor diet quality in this population. This analysis aimed to evaluate associations between those patterns and survival after a cancer diagnosis.

Methods

           We implemented two dietary pattern analysis approaches: penalized logistic regression and principal components analysis. Using nationally representative data from the NHANES (cycles spanning 1999-2018), we extracted six dietary patterns, two of which (the FI and SNAP patterns) were positively associated with being a food-insecure cancer survivor. Cox proportional hazards models were fit to data to assess the relationship between the diet quality indices and survival after a cancer diagnosis.

Results

           There were 981 deaths from all causes, 343 cancer, and 235 cardiovascular disease-related deaths. After multivariable adjustment, we found significantly higher risks of all-cause mortality associated with higher adherence to the FI (HR: 1.23, 95% CI: 1.06-1.42) and SNAP (HR: 1.20, 95% CI: 1.03-1.40) patterns among cancer survivors. Inverse associations were noted between the Prudent #1 pattern and all-cause and cancer-specific mortality. Models with food-insecure cancer survivors revealed non-significant results with parameter estimates commensurate to those from models fit on all cancer survivors.

Conclusion

           Higher adherence to prevailing dietary patterns in the U.S. cancer survivors population may yield a worse prognosis.

**Introduction**

A cancer diagnosis can upend several facets of life and well-being. In addition to psychological distress from the diagnosis, financial toxicity and its accompanying distress can emerge for many cancer survivors owing to exorbitant treatment, prescription, and indirect costs (e.g., lost income due to cancer-related job loss or disability) [1,2]. These phenomena we describe are often magnified for low-income cancer survivors—defined as individuals with a history of cancer—who may lack financial reserves and workplace accommodations while navigating the treatment phases of their cancer [2–4]. Thus, critical questions arise about how these experiences can impact cancer-related outcomes through dietary variables in cancer survivors.

Food insecurity, or the lack of continuous access to healthy and nutritious foods to lead a healthy life, can be a consequence for cancer survivors facing high financial toxicity burdens [5–7]. A framework of competing demands conceptualizes one manifestation of food insecurity amongst cancer survivors, involving cancer survivors facing difficult decisions between choosing medical care or nutritious foods [5]. However, of critical public health concern is that food insecurity is associated with adverse health outcomes and lower diet quality in the general public [8–10]. Food insecurity may predict a worse prognosis among cancer survivors though more research is needed to substantiate this conjecture.

In a previous analysis, we characterized prevailing dietary patterns describing dietary consumption amongst food-insecure cancer survivors using nationally representative data from the National Health and Nutrition Examination Survey (NHANES) [11]. Using penalized logistic regression as a novel supervised learning method for extracting dietary patterns from observed 24-hour recall data, we extracted two major dietary patterns strongly and positively associated with food insecurity among cancer survivors. These patterns emphasized the consumption of added sugars and processed foods and deemphasized fruit and vegetable consumption. Concurrently, we found that a “prudent” dietary pattern emerged that emphasized healthful dietary components such as fruits, vegetables, and whole grains and which was inversely associated with being food-insecure in the same population. Whether these dietary patterns impact clinically meaningful outcomes for cancer survivors was left open-ended.

Understanding how food insecurity impacts different aspects of life, including dietary intake, is a means of delineating at least one potential driving factor behind the health disparities that may arise for cancer survivors experiencing food insecurity. Therefore, this analysis used nationally representative data to examine associations between prevailing dietary patterns in the food-insecure cancer survivor population and the risk of mortality among the broader cancer survivor population and the food-insecure cancer survivor population. The hypothesis was that the dietary patterns describing consumption patterns in the food-insecure cancer survivor population would be positively associated with mortality in the cancer survivor and food-insecure cancer survivor populations.

**Methods**

*Study Setting and Population*

We employed data from ten consecutive cycles (1999-2018) from the NHANES, a biennial cross-sectional study implemented by the Centers for Disease Control and Prevention (CDC) and the National Center for Health Statistics (NCHS) that samples civilian and non-institutionalized community dwellers in the United States. The study implements a complex multi-stage sampling design that generates a nationally representative sample and aims to characterize relationships between lifestyle, medical, physiological, and other factors and health outcomes. It uses surveys that span numerous facets of health and lifestyle and includes a medical examination for a subset of participants. All participants provided informed written consent, and study protocols were approved by the NCHS Ethics Review Board [12]. Because this analysis involved deidentified secondary data, it was exempt from Institutional Review Board approval at the University of Illinois Urbana-Champaign.

In Figure 1, we detail the sample flow that arrives at the final analytical sample size of cancer survivors (*n* = 2493), divided into food-secure participants (*n* = 2176) and food-insecure participants (*n* = 317). Food insecurity status was measured using the United States Department of Agriculture’s (USDA) U.S. Food Security Survey Module (U.S. FSSM), which consists of 18 items designed to evaluate the degree of food insecurity experienced by a participant’s household over the preceding year [13,14]. The survey consists of a series of “yes/no” questions, and responses in the affirmative are used to categorize a household as food-insecure (responding in the affirmative to ≥ 3 items) or food-secure (responding in the affirmative to ≤ 2 items). Cancer history was ascertained via self-report on the Medical Conditions Questionnaire (MCQ). Individuals diagnosed with non-melanoma skin cancer and no other cancer were coded as having no history of cancer, given that the prognosis and benign course of this class of malignancies may otherwise bias the sample [15].

Diagram

Description automatically generated

**Figure 1**. Sample flow diagram detailing inclusion and exclusion criteria for arriving at the final sample.

*Explanatory Variables: Diet Quality Measures*

A thorough description of the computation of the diet quality indices used in the analysis is described elsewhere [11]. Briefly, dietary intake data were amassed by NHANES study staff through two 24-hour recalls using the USDA Automated Multiple-Pass Method—for cycles between 1999-2002, only a single 24-hour recall was performed [16,17]. Nutrient intake data were estimated by referencing the Food and Nutrient Database for Dietary Studies (FNDDS) [18]. Dietary intake and nutrient intake data were averaged across both 24-hour recalls as previously described [11,19,20]. We used the USDA Food Patterns Equivalents Database (FPED) and MyPyramid Equivalents Database (MPED) to obtain intake equivalents of 37 USDA food pattern components, collapsed further into 26 groups, as previously described [11]. Empirical diet quality measures were extracted from observed dietary data with penalized logistic regression (penalized logit) and principal components analysis (PCA). The 26 food groups were the explanatory variables in these models (see Table 2 for the food groups used in this analysis). In the case of the penalized logit models, four binary outcomes were regressed on the centered and scaled transformations of the explanatory variables and included: food insecurity status (food-insecure vs. food-secure), age ≥ 60 years, household receipt of SNAP benefits in the last 12 months, and household size ≥ 5, which are all direct measures, surrogate measures, or risk factors of food insecurity [21,22]. See Maino Vieytes et al. (2022) for an expanded narrative on these procedures and a discussion about the component retention process for the PCA [11].

*Response Variables: All-Cause and Cause-Specific Mortalities*

Mortality and time-to-event data were acquired from the NHANES Public-Use Linked Mortality File, which is generated from deterministic and probabilistic linkages of the NHANES survey data (through the 2017-2018 cycle) to the National Death Index, described elsewhere [23,24]. We computed time since diagnosis and used it as the time scale in our models to minimize potential bias by accounting for left truncation due to delayed enrollment in the study following diagnosis [25–27]. Data were right-censored to either the last known date alive or an administrative censoring date of December 31, 2019. We used the International Classification of Disease, Tenth Revision (ICD-10) codes to classify causes of death. The survival analyses examined all-cause mortality and cause-specific mortality—deaths due to neoplastic malignancy (ICD-10 codes C00-C97) and cardiovascular disease (ICD-10 codes I00-I09, I11, I13, I20-I51, and I60-I69).

*Covariates*

Self-reported demographic and socioeconomic data were obtained in the home interview. Characteristics from the demographic survey (DEMO) included age, sex (*male*/*female*), race and ethnicity (*non-Hispanic White* and *other*), the family income-to-poverty ratio (*< 1.3* or *≥ 1.3*), and household size. We obtained health insurance status (*covered by health insurance*/*not covered by health insurance*) from the health insurance questionnaire (HIQ/HID—for 1999-2004). Behavioral characteristics included smoking status (*current smoker*—currently smoking every day or some days—, *former smoker*—not currently smoking but with a lifetime history of ≥100 cigarettes—, or *never smoker*—a lifetime history of smoking <100 cigarettes), drinking status (*heavy drinker*— ≥ 14 grams/day for women and ≥ 28 grams/day for men—, *moderate drinker*—0.10-13.9 grams/day for women and 0.10-27.9 g/day for men—, and *abstainer*— < 0.10 grams/day), and physical activity (*measured as weekly MET minutes*). These data were obtained from the smoking (SMQ) questionnaire, dietary assessment data, and physical activity questionnaires (PAQ and PAQIAF), respectively [28–30]. Health-related covariates included the Charlson Comorbidity Index score (adapted for NHANES) and body mass index (BMI—kilograms/m2) measured during the physical examination [11,31]. Physical disability was assessed using the 19-item and validated NHANES Activities of Daily Living (ADL) scale, found in the physical functioning questionnaire (PFQ), which is described in detail elsewhere [32,33]. Cancer-related covariates were obtained from the MCQ, and the time since diagnosis was the difference between the age at the time of the survey and the age at the first cancer diagnosis.

*Statistical Analysis*

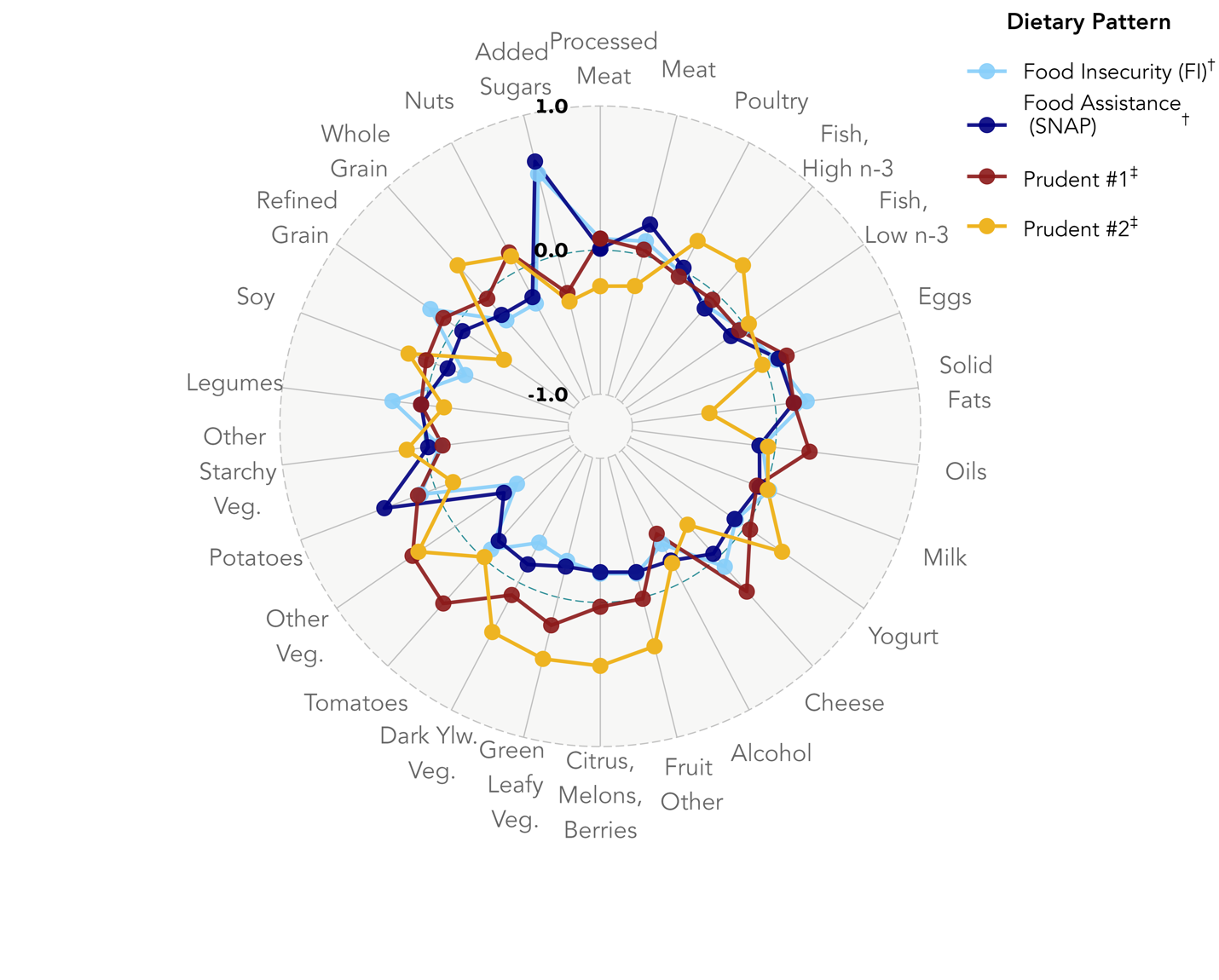
We assessed relationships between diet quality measures and all-cause and cause-specific mortalities with Cox Proportional Hazards models. We implemented several model specifications for the conditional log hazard function to assess the robustness of our results (equations 1-4).

The model in equation 1 specifies the diet quality index, , using 1 dummy variables, , that indicate the subject’s membership in one of the quintiles () of the dietary pattern index score. In equation 2 we conduct a trend test by assigning the subject the median of their respective quintile (where is the set of diet quality index scores for subjects in the quintile) and then modeling as a continuous variable (). In equation 3 we specify the diet index as a continuous variable scaled by the standard deviation of the index and in equation 4 we specify the diet index with a basis expansion of basis functions (not shown here) for a natural cubic spline. Models fit using equation 4 used two interior boundary knots () (not shown here). Given that model 3 is nested in model 4, we used the likelihood ratio test to assess for non-linearity [34,35]. Additionally, all models included terms () for covariates (). We fit these models to data for the entire sample of cancer survivors (*n*  = 2493) and separately on food-insecure cancer survivors (*n* = 317). Model covariates included age, sex, race/ethnicity, BMI, household size, family income-to-poverty ratio, education status, health insurance status, alcohol intake, smoking status, calories, weekly MET minutes, the Charlson Comorbidity Index score, food insecurity status, and the receipt of SNAP benefits [36,37]. Covariates were selected *a priori* based on previous literature and working knowledge about potential confounders in the hypothesized pathway. To account for the possibility of downwardly biased survival estimates from the contributions of participants distantly removed from a cancer diagnosis to the risk set, we conducted a sensitivity analysis including only participants with a primary cancer diagnosis within the five years preceding the date of their interview (*n* = 894). We also considered the NHANES ADL as a covariate, given that food security can be associated with physical disability and functional deficit. However, we did not include it in our primary models, given significant missingness in this variable. Instead, we conducted a subanalysis where we further adjusted for physical disability. All analyses accounted for the complex and probability-based sampling methods of the NHANES study by following the analytical guidelines provided by the NCHS and weighting the analyses accordingly. We used = 0.05 as our threshold level for statistical significance and performed all analyses in R v4.2.2 (The R Foundation, Vienna, Austria). Code and data to reproduce these analyses are publicly accessible at: <https://github.com/cmainov/nhanes-fi-ca-mortality>.

**Results**

The analysis included 603,960 person-months of contributions to the risk set, with 981 deaths from all causes, 343 cancer deaths, and 235 cardiovascular disease-related deaths. The characteristics of the study sample of cancer survivors stratified on food security status are presented in Table 1. On average, food-insecure cancer survivors in this sample were younger than food-secure survivors, more likely to be female, non-White, have a lower educational status, more likely to live under the poverty line, and less likely to be covered by health insurance. Food-insecure cancer survivors were also more likely to live in a home with five or more individuals, be physically or functionally impaired, identify as a current smoker, were less likely to be heavy drinkers, and have a greater comorbidity burden compared to their food-secure counterparts.

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| **Table 1**. Epidemiologic characteristics of the study sample. | | | | |
| **Characteristic** | **Combined Sample** (*n* = 2493) | **Food-insecure** (*n* = 317) | **Food-secure** (*n* = 2176) | ***p*** |
| Age | 62.03 (14.85) | 50.4 (16.46) | 63.32 (14.09) | < 0.01 |
| Sex |  |  |  | < 0.01 |
| Male | 1139 (40.9) | 99 (25.1) | 1040 (42.6) |  |
| Female | 1354 (59.1) | 218 (74.9) | 1136 (57.4) |  |
| Race/Ethnicity |  |  |  | < 0.01 |
| Mexican-American | 174 (2.3) | 51 (8.0) | 123 (1.7) |  |
| Other Hispanic | 133 (2.5) | 40 (7.5) | 93 (1.9) |  |
| Non-Hispanic White | 1730 (86.5) | 156 (70.6) | 1574 (88.3) |  |
| Non-Hispanic Black | 376 (6.2) | 56 (9.2) | 320 (5.9) |  |
| Other/Multiracial | 80 (2.4) | 14 (4.6) | 66 (2.1) |  |
| Education Attained |  |  |  | < 0.01 |
| ≤ High School | 1197 (36.5) | 205 (55.5) | 992 (34.4) |  |
| ≥ Some College | 1296 (63.5) | 112 (44.5) | 1184 (65.6) |  |
| Family Income to Poverty Ratio |  |  |  | < 0.01 |
| < 1.3 | 637 (17.3) | 221 (63.5) | 416 (12.2) |  |
| Health Insurance Status |  |  |  | < 0.01 |
| Insured | 2329 (94.1) | 265 (83.7) | 2064 (95.3) |  |
| Household Size |  |  |  | < 0.01 |
| < 5 Persons | 2274 (92.5) | 247 (78.6) | 2027 (94.1) |  |
| ≥ 5 Persons | 219 (7.5) | 70 (21.4) | 149 (5.9) |  |
| NHANES ADL Score | 22.26 (4.56) | 26.1 (7.34) | 21.93 (4.07) | < 0.01 |
| BMI (kg/m2) | 28.92 (6.61) | 29.82 (7.43) | 28.82 (6.51) | 0.08 |
| Weekly MET Minutes | 2249.04 (4387.81) | 5195.27 (8691.45) | 1923.51 (3462.9) | < 0.01 |
| Daily Caloric Intake (kcal) | 1900.17 (679.88) | 1751 (791.25) | 1916.65 (664.54) | 0.02 |
| Charlson Comorbidity Index | 2.98 (1.35) | 3.36 (1.71) | 2.94 (1.3) | < 0.01 |
| SNAP Benefits |  |  |  | < 0.01 |
| Receiving | 347 (11.2) | 158 (55.6) | 189 (6.3) |  |
| Years Since Diagnosis |  |  |  | 0.62 |
| ≤ 5 years | 894 (32.0) | 119 (30.2) | 775 (32.2) |  |
| > 5 years | 1599 (68.0) | 198 (69.8) | 1401 (67.8) |  |
| Smoking Status |  |  |  | < 0.01 |
| Current | 393 (16.9) | 107 (39.2) | 286 (14.4) |  |
| Former | 1021 (39.4) | 79 (21.9) | 942 (41.3) |  |
| Never | 1079 (43.7) | 131 (38.9) | 948 (44.2) |  |
| Alcohol Use |  |  |  | 0.05 |
| Heavy | 268 (13.8) | 23 (4.6) | 245 (14.9) |  |
| Moderate | 381 (15.9) | 32 (16.1) | 349 (15.9) |  |
| None | 1844 (70.3) | 262 (79.3) | 1582 (69.3) |  |
| Cause of Death |  |  |  | 0.42 |
| Cancer | 343 (36.4) | 30 (31.9) | 313 (36.7) |  |
| Cardiovascular Dis. | 235 (25.6) | 11 (20.4) | 224 (25.9) |  |
| Other | 403 (38.0) | 41 (47.7) | 362 (37.4) |  |
| Percentages may not add to 100% given rounding.  *p-*values are from tests for categorical variables and *t*-tests for continuous variables.  Subjects were weighted, and the analysis was performed according to NCHS guidelines. | | | | |

Table 2 and Figure 2 present weighted Pearson correlation coefficients between the extracted dietary patterns and the individual food groups comprising them. The Food Insecurity (FI) and Food Assistance (SNAP) patterns suggested “unhealthy” dietary intake behavior. Within the sample of all cancer survivors, the FI pattern was characterized by negative correlations with fruits, vegetables, nuts, and whole grains, a high correlation with added sugars, and a weak-to-moderate positive correlation with meat consumption. The SNAP pattern was negatively correlated with fruit and vegetable intake and positively correlated with added sugar consumption, similar to the FI pattern (*r*= 0.81). The pattern of correlation coefficients for the household size pattern was also similar to those from the FI and SNAP patterns and shared a moderate correlation with FI (*r* = 0.63) and SNAP (*r* = 0.60) patterns. Finally, the two patterns extracted with PCA, in general, appeared to reflect “prudent” patterns emphasizing fruit and vegetable intake while de-emphasizing added sugars and were negatively correlated with the FI, SNAP, and Household size patterns. On average, food-insecure subjects had significantly higher scores on the FI and SNAP patterns, with a smaller effect size noted for the household size pattern, and lower scores on the Age, Prudent #1, and Prudent #2 patterns compared to food-secure subjects (Supplementary Table 1).

**Figure 2.** Radar chart with overlay of select dietary patterns and their correlations to the 26 food groups used in the dietary patterns extraction phase of the analysis.

† Dietary pattern obtained using penalized logistic regression; ‡ Dietary pattern obtained using principal components analysis.

Subjects were weighted, and the analysis was performed according to NCHS guidelines.

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| **Table 2**. Pearson correlation coefficients showing the contributions of each food group to the extracted dietary patterns. Correlations amongst the dietary patterns themselves are included at the bottom of the table in the lower triangular matrix form. | | | | | | |
| **Pattern** | **Food Insecurity (FI) †** | **Age †** | **Food Assistance (SNAP) †** | **Household Size†** | **Prudent #1 ‡** | **Prudent #2 ‡** |
| **Food Groups** |
| Processed Meats | 0.01 | -0.01 | 0.08 | 0.05 | 0.08 | -0.25 |
| Meats | 0.22 | -0.04 | 0.10 | 0.02 | 0.04 | -0.22 |
| Poultry | 0.02 | -0.28 | -0.02 | 0.23 | -0.05 | 0.23 |
| Seafood—High n-3 | -0.13 | 0.01 | -0.11 | -0.02 | -0.05 | 0.27 |
| Seafood—Low n-3 | -0.12 | 0.05 | -0.03 | -0.11 | -0.05 | 0.03 |
| Eggs | 0.10 | 0.08 | 0.08 | 0.20 | 0.16 | -0.02 |
| Solid Fats | 0.13 | 0.01 | 0.22 | -0.03 | 0.13 | **-0.46** |
| Oils | -0.11 | -0.01 | -0.09 | 0.15 | 0.24 | -0.05 |
| Milk | -0.04 | **0.34** | 0.03 | -0.13 | -0.06 | 0.02 |
| Yogurt | -0.09 | 0.04 | -0.08 | -0.29 | 0.04 | **0.31** |
| Cheese | -0.04 | **-0.36** | 0.08 | -0.15 | **0.31** | **-0.31** |
| Alcohol | -0.17 | -0.24 | -0.30 | -0.08 | **-0.38** | -0.15 |
| Fruit—Other | -0.18 | **0.35** | -0.17 | -0.27 | 0.01 | **0.35** |
| Fruit—Citrus, melons, and berries | -0.21 | 0.18 | -0.20 | **-0.34** | 0.03 | **0.44** |
| Tomatoes | -0.16 | -0.02 | -0.08 | -0.25 | **0.42** | -0.01 |
| Dark-Green Vegetables | -0.22 | -0.18 | **-0.26** | -0.20 | 0.20 | **0.44** |
| Dark-Yellow Vegetables | -0.14 | 0.09 | **-0.31** | -0.03 | 0.10 | **0.39** |
| Other Vegetables | **-0.41** | 0.10 | **-0.52** | **-0.35** | **0.36** | **0.31** |
| Potatoes | **0.38** | 0.21 | 0.11 | 0.09 | 0.13 | -0.13 |
| Other Starchy Vegetables | -0.02 | 0.16 | -0.10 | -0.13 | -0.12 | 0.13 |
| Legumes | 0.03 | -0.26 | 0.23 | 0.25 | 0.03 | -0.13 |
| Soy | -0.09 | -0.11 | -0.22 | 0.21 | 0.07 | 0.20 |
| Refined Grains | -0.06 | -0.11 | 0.21 | 0.10 | 0.10 | **-0.41** |
| Whole Grains | -0.19 | **0.41** | -0.24 | -0.25 | -0.04 | 0.27 |
| Nuts | -0.21 | 0.09 | -0.26 | 0.00 | 0.14 | 0.11 |
| **Added Sugars** | **0.67** | -0.24 | **0.58** | **0.44** | -0.27 | **-0.33** |
| Food Insecurity (FI)† | -- |  |  |  |  |  |
| Age† | -0.27 | -- |  |  |  |  |
| Food Assistance (SNAP)† | **0.81** | **-0.36** | -- |  |  |  |
| Household Size† | **0.63** | **-0.49** | **0.60** | -- |  |  |
| Prudent #1‡ | -0.24 | 0.09 | **-0.24** | **-0.29** | -- |  |
| Prudent #2‡ | **-0.40** | **0.33** | **-0.57** | **-0.39** | 0.16 | -- |
| † Dietary pattern obtained using penalized logistic regression; ‡ Dietary pattern obtained using principal components analysis. Correlation coefficients (*r*) ≥ |0.30| are bolded to ease the identification of notable food groups characterizing the different patterns. This analysis was performed on all cancer survivors (*n* = 2493).  Subjects were weighted, and the analysis was performed according to NCHS guidelines. | | | | | | |

In our primary analysis and after multivariable adjustment, we found significant associations between the extracted dietary patterns and mortality (Table 3 and Supplementary Table 2). Among the sample of all cancer survivors, the highest quintile of the FI pattern had a 1.52-fold greater risk of all-cause mortality compared to the lowest quintile, and a standard deviation increase in the index score was associated with a 23% increased risk of all-cause mortality. Similarly, the highest quintile of the SNAP pattern score had a 2.17-fold increased risk of all-cause mortality compared to the lowest quintile. A standard deviation increase in the SNAP score was associated with a 1.20-fold greater risk of all-cause mortality. Figure 3 presents survival curves and spline curves for these relationships. The parameter estimates were similar when we performed the analysis on food-insecure cancer survivors, albeit they had higher variance (Supplementary Table 2).

In contrast, inverse associations were noted for the two “prudent” patterns extracted via PCA. Amongst all cancer survivors, the highest quintile of Prudent pattern #1 had a 46% decreased risk of all-cause mortality compared to the lowest quintile and a 20% decreased risk associated with a standard deviation increase in the score. Within food-insecure cancer survivors, the highest quintile of Prudent pattern #2 had an 82% reduced risk of all-cause mortality compared to the first quintile with a significant test for linear trend.

When we examined cancer-specific mortality, the parameter estimates amongst all cancer survivors were similar to those for all-cause mortality, particularly for the FI pattern (Table 3). However, no results other than an inverse association involving Prudent pattern #1 were statistically significant at the level of = 0.05. Considering cardiovascular disease mortality, the effect sizes were close to null, and we observed a significant and inverse association between Prudent pattern #1 and the risk of cardiovascular disease-related mortality. Further adjusting for the NHANES ADL score did not significantly alter the results (Supplementary Table 3) despite losing many participants from the risk set. Finally, in our sensitivity analysis that included only participants with a primary cancer diagnosis within the five years before their study interview (Supplementary Table 4), we found that the association between the FI pattern and all-cause mortality were slightly amplified. The SNAP pattern in this analysis yielded similar results as those observed in the primary analysis. Notably, relationships between Prudent patterns #1 and #2 and all-cause mortality attenuated towards the null.

|  |  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- | --- |
| **Table 3**. Adjusted hazard ratios and 95% confidence intervals for the risks of all-cause and cause-specific mortalities, in relation to the dietary patterns, in the cancer survivor sample (*n* = 2493). | | | | | | | | | |
| **Dietary Pattern**d | ***n*** | **Q1** | **Q2** | **Q3** | **Q4** | **Q5** | ***p***a**trend** | **HR**b**continuous** | ***p***c**non-linear** |
| *All-Cause Mortality* | | | | | | | | | |
| Food Insecurity (FI)**†** | 2493 | 1.00 | 0.91 (0.66-1.24) | 0.90 (0.65-1.25) | 1.05 (0.72-1.53) | 1.52 (1.01-2.29)\* | 0.03\* | 1.23 (1.06-1.42)\*\* | 0.48 |
| Age**†** | 2493 | 1.00 | 0.85 (0.57-1.27) | 1.05 (0.67-1.65) | 0.94 (0.64-1.40) | 1.13 (0.74-1.72) | 0.41 | 1.08 (0.94-1.24) | 0.82 |
| Household Size**†** | 2493 | 1.00 | 0.99 (0.75-1.32) | 1.05 (0.76-1.45) | 1.05 (0.76-1.45) | 1.17 (0.74-1.86) | 0.46 | 1.06 (0.91-1.23) | 0.97 |
| Food Assistance (SNAP) **†** | 2493 | 1.00 | 1.21 (0.90-1.62) | 1.05 (0.76-1.44) | 1.25 (0.90-1.73) | 2.17 (1.45-3.24)\*\* | < 0.01\*\* | 1.20 (1.03-1.40)\* | 0.42 |
| Prudent #1**‡** | 2493 | 1.00 | 0.92 (0.68-1.25) | 0.95 (0.69-1.32) | 0.68 (0.46-1.00) | 0.54 (0.36-0.82)\*\* | < 0.01\*\* | 0.80 (0.70-0.92)\*\* | 0.02\* |
| Prudent #2**‡** | 2493 | 1.00 | 1.32 (0.88-1.99) | 0.92 (0.61-1.40) | 0.96 (0.63-1.46) | 0.94 (0.61-1.44) | 0.34 | 0.97 (0.86-1.09) | 0.70 |
| *Cancer-Specific Mortality* | | | | | | | | | |
| Food Insecurity (FI)**†** | 2493 | 1.00 | 0.83 (0.50-1.37) | 0.93 (0.58-1.49) | 0.93 (0.55-1.58) | 1.49 (0.87-2.57) | 0.13 | 1.23 (0.99-1.51) | 0.75 |
| Age**†** | 2493 | 1.00 | 0.64 (0.37-1.11) | 0.69 (0.36-1.29) | 0.88 (0.53-1.47) | 0.62 (0.35-1.10) | 0.30 | 0.92 (0.76-1.11) | 0.93 |
| Household Size**†** | 2493 | 1.00 | 0.84 (0.49-1.43) | 0.97 (0.62-1.52) | 0.76 (0.48-1.20) | 1.00 (0.53-1.91) | 0.89 | 1.04 (0.86-1.26) | 0.63 |
| Food Assistance (SNAP) **†** | 2493 | 1.00 | 0.87 (0.57-1.34) | 0.98 (0.63-1.51) | 1.02 (0.60-1.71) | 1.80 (0.92-3.51) | 0.12 | 1.17 (0.92-1.48) | 0.38 |
| Prudent #1**‡** | 2493 | 1.00 | 0.86 (0.52-1.44) | 0.91 (0.56-1.48) | 0.54 (0.29-0.98)\* | 0.71 (0.41-1.25) | 0.12 | 0.83 (0.70-0.98)\* | 0.57 |
| Prudent #2**‡** | 2493 | 1.00 | 1.38 (0.75-2.52) | 1.01 (0.55-1.86) | 0.68 (0.35-1.35) | 1.03 (0.54-1.96) | 0.49 | 0.99 (0.82-1.21) | 0.49 |
| *Cardiovascular Disease Mortality* | | | | | | | | | |
| Food Insecurity (FI)**†** | 2493 | 1.00 | 0.84 (0.41-1.72) | 1.08 (0.53-2.20) | 1.33 (0.65-2.74) | 1.05 (0.47-2.37) | 0.51 | 1.16 (0.89-1.50) | 0.77 |
| Age**†** | 2493 | 1.00 | 1.48 (0.69-3.18) | 2.23 (1.01-4.92)\* | 1.67 (0.88-3.18) | 1.72 (0.93-3.18) | 0.22 | 1.19 (0.98-1.43) | 0.72 |
| Household Size**†** | 2493 | 1.00 | 1.00 (0.56-1.80) | 1.73 (0.99-3.02) | 1.67 (0.86-3.27) | 1.20 (0.60-2.41) | 0.15 | 1.13 (0.94-1.37) | 0.08 |
| Food Assistance (SNAP) **†** | 2493 | 1.00 | 1.34 (0.69-2.59) | 1.36 (0.77-2.40) | 1.12 (0.62-2.03) | 2.08 (1.13-3.82)\* | 0.12 | 1.14 (0.91-1.43) | 0.94 |
| Prudent #1**‡** | 2493 | 1.00 | 1.12 (0.60-2.08) | 0.86 (0.40-1.83) | 0.54 (0.28-1.06) | 0.40 (0.18-0.88)\* | < 0.01\*\* | 0.70 (0.57-0.87)\*\* | 0.09 |
| Prudent #2**‡** | 2493 | 1.00 | 0.80 (0.41-1.57) | 1.01 (0.52-1.96) | 0.92 (0.45-1.85) | 1.21 (0.54-2.71) | 0.50 | 1.09 (0.88-1.33) | 0.53 |
| \*\* *p* < 0.01; \* *p* < 0.05  a Test for trend across the quintiles of the dietary exposure. See Equation 2 in the main text.  b Hazard ratio for a standard deviation increase in the dietary exposure. See Equation 3 in the main text.  c Likelihood ratio test *p*-value for a natural cubic spline model (Equation 4 in the main text) compared to specifying the model with the scaled dietary exposure (Equation 3).  d All models adjusted for age, sex, race, BMI, household size, family income-to-poverty ratio, education status, health insurance status, receipt of SNAP benefits, food insecurity status, alcohol intake, smoking status, total caloric intake, weekly MET minutes, and the Charlson Comorbidity Index score.  **†** Dietary pattern obtained using penalized logistic regression; **‡** Dietary pattern obtained using principal components analysis (PCA).  Subjects were weighted, and the analysis was performed according to NCHS guidelines. | | | | | | | | | |

A picture containing text, diagram, plot, line

Description automatically generated

**Figure 3**. Relationships between the FI (panels A and B) and SNAP (panels C and D) patterns and all-cause mortality in cancer survivors (*n* = 2493) categorized into quintiles, with quintile 1 having the lowest dietary pattern scores and quintile 5 having the highest. Adjusted survival curves were generated from models specified with quintile dummy variables and spline curves from expanding the diet quality index using a basis expansion for a natural cubic spline with four interior knots. These models adjusted for age, sex, race, BMI, household size, family income-to-poverty ratio, education status, health insurance status, alcohol intake, smoking status, calories, weekly MET minutes, the Charlson Comorbidity Index score, receipt of SNAP benefits, and food insecurity status*.*

aDietary pattern scores were normalized prior to plotting

Subjects were weighted, and the analysis was performed according to NCHS guidelines.

**Discussion**

Using a nationally representative sample of U.S. cancer survivors, we found that dietary patterns associated with being a food-insecure cancer survivor were positively associated with all-cause and cancer-specific mortality after adjusting for several confounders. In a previous analysis, we validated the utility of penalized logistic regression as a novel *a posteriori* method to extract dietary patterns associated with a particular risk factor or outcome [11]. The results of this follow-up analysis demonstrate the clinical value of these dietary patterns and their relationships to survival. Of the six dietary patterns we extracted from the observed 24-hour recall data (four with penalized logit and two with PCA), two of these patterns—the FI and SNAP patterns, which were both loaded by high consumption of palatable and processed foods and low loadings of fruits, vegetables, and other healthful components—were robustly and positively associated with all-cause and cancer-specific mortalities amongst cancer survivors and a subset of food-insecure cancer survivors. There was also evidence that the prudent-style patterns (i.e., those characterized by high consumption of fruits, vegetables, and other healthy components) extracted with PCA, which were inversely correlated with food insecurity status, were also inversely associated with all-cause and cancer-specific mortality. However, the strength of the evidence for these patterns was not as strong as for the others mentioned. Moreover, the results we observed were robust after performing multiple sensitivity analyses.

Our findings contribute to evidence highlighting adverse associations between food insecurity and health outcomes. However, our work is novel in that we focused on cancer survivors, a population that has, overall, received relatively little scrutiny within the broader context of food insecurity, despite that this population may have an elevated risk of experiencing food insecurity [5,38,39]. Several lines of evidence tie food insecurity to an increased comorbidity burden, including increased risks of hypertension, hyperlipidemia, diabetes, and mental health conditions [8,40–42]. Moreover, food insecurity is associated with poor overall health status, and recent analyses using the NHANES data demonstrated significant and positive associations between food insecurity status and the risk of all-cause and cardiovascular disease mortality [21,37,41,43,44]. Our analysis complements this previous body of work by demonstrating that dietary intake may be pertinent to the pathway between food insecurity and increased mortality in cancer survivors.

Across studies, place, populations, and time, diet quality is associated with physiological outcomes that may explain the differential propensity for survival like those observed in our analysis. In a longitudinal sample of older adults from the Health and Retirement Study, higher diet quality, as measured by the HEI-2015, was associated with better lipid and C-reactive protein (CRP) profiles and decreased likelihood of depression and functional deficits [45]. In another longitudinal sample from the Health, Eating, Activity, and Lifestyle (HEAL) prospective cohort study, higher postdiagnosis HEI-2015 scores were associated with lower CRP levels in a sample of breast-cancer survivors [46]. A nested cross-sectional study from the Multiethnic Cohort Study examined relationships between four *a priori* diet quality indices (AHEI-2010, HEI-2010, aMED, and DASH) and several serum carotenoids and biomarkers (leptin, HOMA-IR, glucose, CRP, insulin and triglycerides) and found that higher diet index scores were positively associated with the carotenoid markers and inversely associated with the other biomarkers [47]. Finally, in a cross-sectional analysis of newly diagnosed head and neck squamous cell carcinoma patients, higher diet quality, measured by a “whole foods” dietary pattern extracted using PCA from FFQ data, was inversely associated with several pro-inflammatory cytokines [48]. Thus, the link between diet quality and downstream inflammation may explain our observed results, particularly in the context of cancer, where higher inflammatory biomarkers exacerbate disease progression resulting in poor prognosis [49–54].

Our findings have important policy implications. As we alluded to in our previous analysis, screening for food insecurity is not a clinical best practice widely implemented in cancer clinics, though the National Comprehensive Cancer Network recently incorporated an item dedicated to food insecurity in its Distress Thermometer screener [5,11,55]. Alongside calls from those also working in this area, we believe our results are grounds for further pushing food insecurity and cancer survivorship to the forefront of discussion within major medical organizations [56]. Identifying food-insecure patients in the oncology care setting, early on in the cancer care continuum and later on as care progresses and financial hardship may be exacerbated, can facilitate prompt referral to additional support resources. These resources could be accessed, for example, through a case manager or social worker that assists the cancer survivor to leverage personal and community-level resources or provides referrals to federal and local nutrition assistance programs [5,57]. The establishment of hospital-based food pantries is another avenue that has shown promise for cancer survivors to access nutritious foods they may otherwise lack access to [49]. Thus, tailoring community and higher-level initiatives prioritizing food support throughout the treatment phase and the prolonged post-treatment phase may be critical to mitigating the negative health consequences food-insecure cancer survivors may experience secondary to a lack of a steady stream of nutritious foods [59]. Though community-level strategies may yield benefits, particularly in the short-term, more comprehensive systems-level approaches, such as innovative medical insurance models, are needed for more wide-ranging effects on food-insecure survivors [5,60,61].

This analysis has several strengths, including the nationally representative sample from the NHANES, the use of a validated food insecurity measurement tool, the quality and quantity of covariate data we used to account for potential and known confounders, and the quality of the linked mortality data through the NCHS Linked Mortality Files. Nevertheless, there are limitations to note. First, we used a static measure of dietary intake though we know that patterns of dietary intake are dynamic and circumstantial, and our analysis was not able to account for any variation in dietary intake across time despite using time-to-event measures that occurred substantially after the measurement instance. In a similar vein, it is worth considering that food insecurity can be a transient phenomenon that subjects recover from, and this may have occurred for participants in the intervening window between the study visit and the time of the observed event or censoring. An additional consideration concerning the measurement of dietary intake using 24-hour recalls is that it may be subject to systematic measurement errors that we were not able to quantify with the available data. Second, with any analysis of observational data outside of a rigid set of assumptions, we must conclude that unmeasured or residual confounding cannot be excluded and that no causal interpretations can be made with these results. Third, though we did not account for stress as a confounding variable, given limitations with measures of psychological stress or allostatic load on the NHANES survey and our sample size, we believe it is appropriate to conclude that measures of food insecurity, such as those captured by the USDA FSSM, are likely to be highly correlated with measures of stress. Indeed, the U.S. Household FSSM includes questions designed to capture concern and stress about food insufficiency, such as:*“(I/We) worried whether (my/our) food would run out before (I/we) got money to buy more”* [14]. Finally, a critical reflection concerning the use of the U.S. Household FSSM is that a measure of household food insecurity may not capture the burden of food insecurity exacted on any one individual within that household.

In summary, we conclude that major and prevailing dietary patterns amongst U.S. food-insecure cancer survivors, derived empirically using a novel supervised learning methodology, may deleteriously impact cancer-related outcomes such as all-cause and cause-specific mortality. These patterns, characterized by the consumption of added sugars and processed foods with concomitant low consumption of fruits, vegetables, whole grains, and other healthful diet components, and our findings signal an urgent public health challenge demanding innovative policy and community-level solutions. We identify several avenues of future research in this area. One avenue includes developing and evaluating community and individual-level interventions for bolstering food security amongst food-insecure cancer survivors throughout the early treatment and cost-prohibitive phases of the cancer-care continuum. A second avenue should focus on piloting interventions for medical provider training on screening for food insecurity in oncology settings. A final avenue of research should extend our work and continue surveillance of dietary intake patterns amongst U.S. food-insecure cancer survivors using nationally representative data. Ultimately, advances in such areas will ideally abate the disparities in health outcomes observed by food-insecure cancer survivors that our work and others highlight.

**Conflict of Interest**

The authors declare no potential conflicts of interest.

**Funding**

CAMV is a scholar in Health Policy Research Scholars, a national leadership program supported by the Robert Wood Johnson Foundation. It supports scholars from diverse disciplines and backgrounds in applying and advocating for policy change that improves health and equity. The project described in this article is supported by the program. The views expressed here do not necessarily reflect the views of the Foundation. [www.healthpolicyresearch-scholars.org](http://www.healthpolicyresearch-scholars.org).

**References**

1. Yousuf Zafar, S. Financial Toxicity of Cancer Care: It’s Time to Intervene. *JNCI J. Natl. Cancer Inst.* **2016**, *108*, djv370, doi:10.1093/jnci/djv370.

2. Carrera, P.M.; Kantarjian, H.M.; Blinder, V.S. The Financial Burden and Distress of Patients with Cancer: Understanding and Stepping-up Action on the Financial Toxicity of Cancer Treatment. *CA. Cancer J. Clin.* **2018**, *68*, 153–165, doi:10.3322/caac.21443.

3. Blinder, V.; Eberle, C.; Patil, S.; Gany, F.M.; Bradley, C.J. Women With Breast Cancer Who Work For Accommodating Employers More Likely To Retain Jobs After Treatment. *Health Aff. (Millwood)* **2017**, *36*, 274–281, doi:10.1377/hlthaff.2016.1196.

4. Cancer Survivor.

5. Patel, K.G.; Borno, H.T.; Seligman, H.K. Food Insecurity Screening: A Missing Piece in Cancer Management. *Cancer* **2019**, *125*, 3494–3501, doi:10.1002/cncr.32291.

6. Charkhchi, P.; Fazeli Dehkordy, S.; Carlos, R.C. Housing and Food Insecurity, Care Access, and Health Status Among the Chronically Ill: An Analysis of the Behavioral Risk Factor Surveillance System. *J. Gen. Intern. Med.* **2018**, *33*, 644–650, doi:10.1007/s11606-017-4255-z.

7. Food Security in the U.S. 2023.

8. Seligman, H.K.; Schillinger, D. Hunger and Socioeconomic Disparities in Chronic Disease. *N. Engl. J. Med.* **2010**, *363*, 6–9, doi:10.1056/NEJMp1000072.

9. Gundersen, C.; Ziliak, J.P. Food Insecurity And Health Outcomes. *Health Aff. (Millwood)* **2015**, *34*, 1830–1839, doi:10.1377/hlthaff.2015.0645.

10. Leung, C.W.; Epel, E.S.; Ritchie, L.D.; Crawford, P.B.; Laraia, B.A. Food Insecurity Is Inversely Associated with Diet Quality of Lower-Income Adults. *J. Acad. Nutr. Diet.* **2014**, *114*, 1943-1953.e2, doi:10.1016/j.jand.2014.06.353.

11. Maino Vieytes, C.A.; Zhu, R.; Gany, F.; Burton-Obanla, A.; Arthur, A.E. Empirical Dietary Patterns Associated with Food Insecurity in U.S. Cancer Survivors: NHANES 1999–2018. *Int. J. Environ. Res. Public. Health* **2022**, *19*, 14062, doi:10.3390/ijerph192114062.

12. About the National Health and Nutrition Examination Survey.

13. Bickel, G.; Nord, M.; Price, C.; Hamilton, W.; Cook, J. Guide to Measuring Household Food Security 2000.

14. U.S. Household Food Insecurity Survey Module: Three-Stage Design, With Screeners.

15. Yaghjyan, L.; Wijayabahu, A.T.; Egan, K.M. RE: The Association Between Dietary Quality and Overall and Cancer-Specific Mortality Among Cancer Survivors, NHANES III. *JNCI Cancer Spectr.* **2018**, *2*, pky044, doi:10.1093/jncics/pky044.

16. Blanton, C.A.; Moshfegh, A.J.; Baer, D.J.; Kretsch, M.J. The USDA Automated Multiple-Pass Method Accurately Estimates Group Total Energy and Nutrient Intake. *J. Nutr.* **2006**, *136*, 2594–2599, doi:10.1093/jn/136.10.2594.

17. Moshfegh, A.J.; Rhodes, D.G.; Baer, D.J.; Murayi, T.; Clemens, J.C.; Rumpler, W.V.; Paul, D.R.; Sebastian, R.S.; Kuczynski, K.J.; Ingwersen, L.A.; et al. The US Department of Agriculture Automated Multiple-Pass Method Reduces Bias in the Collection of Energy Intakes. *Am. J. Clin. Nutr.* **2008**, *88*, 324–332, doi:10.1093/ajcn/88.2.324.

18. Montville, J.B.; Ahuja, J.K.C.; Martin, C.L.; Heendeniya, K.Y.; Omolewa-Tomobi, G.; Steinfeldt, L.C.; Anand, J.; Adler, M.E.; LaComb, R.P.; Moshfegh, A. USDA Food and Nutrient Database for Dietary Studies (FNDDS), 5.0. *Procedia Food Sci.* **2013**, *2*, 99–112, doi:10.1016/j.profoo.2013.04.016.

19. Jovanovic, C.E.S.; Hoelscher, D.M.; Chen, B.; Ranjit, N.; van den Berg, A.E. The Associations of Plant-Based Food and Metabolic Syndrome Using NHANES 2015–16 Data. *J. Public Health* **2022**, fdab403, doi:10.1093/pubmed/fdab403.

20. Moore, C.; Murphy, M.M.; Keast, D.R.; Holick, M.F. Vitamin D Intake in the United States. *J. Am. Diet. Assoc.* **2004**, *104*, 980–983, doi:10.1016/j.jada.2004.03.028.

21. Lee, J.S.; Frongillo, E.A. Nutritional and Health Consequences Are Associated with Food Insecurity among U.S. Elderly Persons. *J. Nutr.* **2001**, *131*, 1503–1509, doi:10.1093/jn/131.5.1503.

22. Kohn, M.J.; Bell, J.F.; Grow, H.M.G.; Chan, G. Food Insecurity, Food Assistance and Weight Status in US Youth: New Evidence from NHANES 2007-08: Food Insecurity, Assistance and Weight. *Pediatr. Obes.* **2014**, *9*, 155–166, doi:10.1111/j.2047-6310.2012.00143.x.

23. Linkage Methods and Analytical Support for NCHS Linked Mortality Data.

24. Continuous NHANES Public-Use Linked Mortality Files, 2019. Hyattsville, Maryland. *Natl. Cent. Health Stat. Div. Anal. Epidemiol.*, doi:10.15620/cdc:117142.

25. Cain, K.C.; Harlow, S.D.; Little, R.J.; Nan, B.; Yosef, M.; Taffe, J.R.; Elliott, M.R. Bias Due to Left Truncation and Left Censoring in Longitudinal Studies of Developmental and Disease Processes. *Am. J. Epidemiol.* **2011**, *173*, 1078–1084, doi:10.1093/aje/kwq481.

26. Foreman, A.; Lai, G.; Miller, D. Surviving Left Truncation Using PROC PHREG.

27. Song, M.; Wu, K.; Meyerhardt, J.A.; Yilmaz, O.; Wang, M.; Ogino, S.; Fuchs, C.S.; Giovannucci, E.L.; Chan, A.T. Low-Carbohydrate Diet Score and Macronutrient Intake in Relation to Survival After Colorectal Cancer Diagnosis. *JNCI Cancer Spectr.* **2018**, *2*, pky077, doi:10.1093/jncics/pky077.

28. Agarwal, S. The Association of Active and Passive Smoking with Peripheral Arterial Disease: Results from NHANES 1999–2004. *Angiology* **2009**, *60*, 335–345, doi:10.1177/0003319708330526.

29. Dietary Guidelines Advisory Committee; OverDrive, I. *Dietary Guidelines for Americans 2015-2020*; 2016; ISBN 978-0-16-093465-0.

30. Tucker, L.A. Physical Activity and Telomere Length in U.S. Men and Women: An NHANES Investigation. *Prev. Med.* **2017**, *100*, 145–151, doi:10.1016/j.ypmed.2017.04.027.

31. Zhao, H.; Pan, Y.; Wang, C.; Guo, Y.; Yao, N.; Wang, H.; Li, B. The Effects of Metal Exposures on Charlson Comorbidity Index Using Zero-Inflated Negative Binomial Regression Model: NHANES 2011–2016. *Biol. Trace Elem. Res.* **2021**, *199*, 2104–2111, doi:10.1007/s12011-020-02331-4.

32. Cook, C.E.; Richardson, J.K.; Pietrobon, R.; Braga, L.; Silva, H.M.; Turner, D. Validation of the NHANES ADL Scale in a Sample of Patients with Report of Cervical Pain: Factor Analysis, Item Response Theory Analysis, and Line Item Validity. *Disabil. Rehabil.* **2006**, *28*, 929–935, doi:10.1080/09638280500404263.

33. Kalyani, R.R.; Saudek, C.D.; Brancati, F.L.; Selvin, E. Association of Diabetes, Comorbidities, and A1C With Functional Disability in Older Adults. *Diabetes Care* **2010**, *33*, 1055–1060, doi:10.2337/dc09-1597.

34. Witte, J.S.; Greenland, S. A Nested Approach to Evaluating Dose-Response and Trend. *Ann. Epidemiol.* **1997**, *7*, 188–193, doi:10.1016/S1047-2797(96)00159-7.

35. Woolf, B. The Log Likelihood Ratio Test (The G-Test). *Ann. Hum. Genet.* **1957**, *21*, 397–409, doi:10.1111/j.1469-1809.1972.tb00293.x.

36. Willett, W.C.; Howe, G.R.; Kushi, L.H. Adjustment for Total Energy Intake in Epidemiologic Studies. *Am. J. Clin. Nutr.* **1997**, *65*, 1220S-1228S, doi:10.1093/ajcn/65.4.1220S.

37. Banerjee, S.; Radak, T.; Khubchandani, J.; Dunn, P. Food Insecurity and Mortality in American Adults: Results From the NHANES-Linked Mortality Study. *Health Promot. Pract.* **2021**, *22*, 204–214, doi:10.1177/1524839920945927.

38. Gany, F.; Lee, T.; Ramirez, J.; Massie, D.; Moran, A.; Crist, M.; McNish, T.; Winkel, G.; Leng, J.C.F. Do Our Patients Have Enough to Eat?: Food Insecurity among Urban Low-Income Cancer Patients. *J. Health Care Poor Underserved* **2014**, *25*, 1153–1168, doi:10.1353/hpu.2014.0145.

39. Robien, K.; Clausen, M.; Sullo, E.; Ford, Y.R.; Griffith, K.A.; Le, D.; Wickersham, K.E.; Wallington, S.F. Prevalence of Food Insecurity Among Cancer Survivors in the United States: A Scoping Review. *J. Acad. Nutr. Diet.* **2023**, *123*, 330–346, doi:10.1016/j.jand.2022.07.004.

40. Seligman, H.K.; Bindman, A.B.; Vittinghoff, E.; Kanaya, A.M.; Kushel, M.B. Food Insecurity Is Associated with Diabetes Mellitus: Results from the National Health Examination and Nutrition Examination Survey (NHANES) 1999-2002. *J. Gen. Intern. Med.* **2007**, *22*, 1018–1023, doi:10.1007/s11606-007-0192-6.

41. Stuff, J.E.; Casey, P.H.; Szeto, K.L.; Gossett, J.M.; Robbins, J.M.; Simpson, P.M.; Connell, C.; Bogle, M.L. Household Food Insecurity Is Associated with Adult Health Status. *J. Nutr.* **2004**, *134*, 2330–2335, doi:10.1093/jn/134.9.2330.

42. Heflin, C.M.; Ziliak, J.P. Food Insufficiency, Food Stamp Participation, and Mental Health\*. *Soc. Sci. Q.* **2008**, *89*, 706–727, doi:10.1111/j.1540-6237.2008.00556.x.

43. Sun, Y.; Liu, B.; Rong, S.; Du, Y.; Xu, G.; Snetselaar, L.G.; Wallace, R.B.; Bao, W. Food Insecurity Is Associated With Cardiovascular and All‐Cause Mortality Among Adults in the United States. *J. Am. Heart Assoc.* **2020**, *9*, doi:10.1161/JAHA.119.014629.

44. Vozoris, N.T.; Tarasuk, V.S. Household Food Insufficiency Is Associated with Poorer Health. *J. Nutr.* **2003**, *133*, 120–126, doi:10.1093/jn/133.1.120.

45. Zhao, H.; Andreyeva, T. Diet Quality and Health in Older Americans. *Nutrients* **2022**, *14*, 1198, doi:10.3390/nu14061198.

46. George, S.M.; Neuhouser, M.L.; Mayne, S.T.; Irwin, M.L.; Albanes, D.; Gail, M.H.; Alfano, C.M.; Bernstein, L.; McTiernan, A.; Reedy, J.; et al. Postdiagnosis Diet Quality Is Inversely Related to a Biomarker of Inflammation among Breast Cancer Survivors. *Cancer Epidemiol. Biomarkers Prev.* **2010**, *19*, 2220–2228, doi:10.1158/1055-9965.EPI-10-0464.

47. Guillermo, C.; Boushey, C.J.; Franke, A.A.; Monroe, K.R.; Lim, U.; Wilkens, L.R.; Marchand, L.L.; Maskarinec, G. Diet Quality and Biomarker Profiles Related to Chronic Disease Prevention: The Multiethnic Cohort Study. *J. Am. Coll. Nutr.* **2020**, *39*, 216–223, doi:10.1080/07315724.2019.1635921.

48. Arthur, A.E.; Peterson, K.E.; Shen, J.; Djuric, Z.; Taylor, J.M.G.; Hebert, J.R.; Duffy, S.A.; Peterson, L.A.; Bellile, E.L.; Whitfield, J.R.; et al. Diet and Proinflammatory Cytokine Levels in Head and Neck Squamous Cell Carcinoma. *Cancer* **2014**, *120*, 2704–2712, doi:10.1002/cncr.28778.

49. Brenner, D.R.; Scherer, D.; Muir, K.; Schildkraut, J.; Boffetta, P.; Spitz, M.R.; Le Marchand, L.; Chan, A.T.; Goode, E.L.; Ulrich, C.M.; et al. A Review of the Application of Inflammatory Biomarkers in Epidemiologic Cancer Research. *Cancer Epidemiol. Biomarkers Prev.* **2014**, *23*, 1729–1751, doi:10.1158/1055-9965.EPI-14-0064.

50. Cheng, E.; Shi, Q.; Shields, A.F.; Nixon, A.B.; Shergill, A.P.; Ma, C.; Guthrie, K.A.; Couture, F.; Kuebler, P.; Kumar, P.; et al. Association of Inflammatory Biomarkers With Survival Among Patients With Stage III Colon Cancer. *JAMA Oncol.* **2023**, doi:10.1001/jamaoncol.2022.6911.

51. Domenici, L.; Tonacci, A.; Aretini, P.; Garibaldi, S.; Perutelli, A.; Bottone, P.; Muzii, L.; Panici, P.B. Inflammatory Biomarkers as Promising Predictors of Prognosis in Cervical Cancer Patients. *Oncology* **2021**, *99*, 571–579, doi:10.1159/000517320.

52. Petekkaya, I.; Unlu, O.; Roach, E.C.; Gecmez, G.; Okoh, A.K.; Babacan, T.; Sarici, F.; Keskin, O.; Arslan, C.; Petekkaya, E.; et al. Prognostic Role of Inflammatory Biomarkers in Metastatic Breast Cancer.

53. Nelson, S.H.; Brasky, T.M.; Patterson, R.E.; Laughlin, G.A.; Kritz-Silverstein, D.; Edwards, B.J.; Lane, D.; Rohan, T.E.; Ho, G.Y.F.; Manson, J.E.; et al. The Association of the C-Reactive Protein Inflammatory Biomarker with Breast Cancer Incidence and Mortality in the Women’s Health Initiative. *Cancer Epidemiol. Biomark. Prev. Publ. Am. Assoc. Cancer Res. Cosponsored Am. Soc. Prev. Oncol.* **2017**, *26*, 1100–1106, doi:10.1158/1055-9965.EPI-16-1005.

54. Pierce, B.L.; Ballard-Barbash, R.; Bernstein, L.; Baumgartner, R.N.; Neuhouser, M.L.; Wener, M.H.; Baumgartner, K.B.; Gilliland, F.D.; Sorensen, B.E.; McTiernan, A.; et al. Elevated Biomarkers of Inflammation Are Associated With Reduced Survival Among Breast Cancer Patients. *J. Clin. Oncol.* **2009**, *27*, 3437–3444, doi:10.1200/JCO.2008.18.9068.

55. Distress Thermometer Tool Translations.

56. Burton-Obanla, A.A.; Sloane, S.; Koester, B.; Gundersen, C.; Fiese, B.H.; Arthur, A.E. Oncology Registered Dietitian Nutritionists’ Knowledge, Attitudes, and Practices Related to Food Insecurity among Cancer Survivors: A Qualitative Study. *J. Acad. Nutr. Diet.* **2022**, *122*, 2267–2287, doi:10.1016/j.jand.2021.12.004.

57. Seligman, H.K.; Berkowitz, S.A. Aligning Programs and Policies to Support Food Security and Public Health Goals in the United States. *Annu. Rev. Public Health* **2019**, *40*, 319–337, doi:10.1146/annurev-publhealth-040218-044132.

58. Gany, F.; Lee, T.; Loeb, R.; Ramirez, J.; Moran, A.; Crist, M.; McNish, T.; Leng, J.C.F. Use of Hospital-Based Food Pantries Among Low-Income Urban Cancer Patients. *J. Community Health* **2015**, *40*, 1193–1200, doi:10.1007/s10900-015-0048-7.

59. Gany, F.; Melnic, I.; Wu, M.; Li, Y.; Finik, J.; Ramirez, J.; Blinder, V.; Kemeny, M.; Guevara, E.; Hwang, C.; et al. Food to Overcome Outcomes Disparities: A Randomized Controlled Trial of Food Insecurity Interventions to Improve Cancer Outcomes. *J. Clin. Oncol.* **2022**, *40*, 3603–3612, doi:10.1200/JCO.21.02400.

60. Berkowitz, S.A.; Seligman, H.K.; Choudhry, N.K. Treat or Eat: Food Insecurity, Cost-Related Medication Underuse, and Unmet Needs. *Am. J. Med.* **2014**, *127*, 303-310.e3, doi:10.1016/j.amjmed.2014.01.002.

61. Moellman, N. Healthcare and Hunger: Effects of the ACA Medicaid Expansions on Food Insecurity in America. *Appl. Econ. Perspect. Policy* **2020**, *42*, 168–186, doi:10.1093/aepp/ppz018.