

# Examining the Impact of Macronutrient Composition on Caloric Value: A Structural Equation Modeling (SEM) Approach

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## ABSTRACT

This study examined the impact of macronutrient composition on the caloric value of food items using a Structural Equation Modeling (SEM) approach. The research aimed to fill a gap in the literature by analyzing the combined effects of carbohydrates, proteins, and fats on caloric content, rather than focusing on individual macronutrients in isolation. Data were sourced from the Food Nutrition Dataset, which provided detailed nutritional information on a wide range of food items. The SEM analysis revealed that fats had the largest impact on caloric value, followed by proteins and carbohydrates. The study found that while fats contributed significantly to the overall caloric content, proteins and carbohydrates also played crucial roles. The findings underscored the importance of considering the synergistic effects of macronutrients when estimating caloric value. Practical applications of this research included aiding dietitians in designing balanced meal plans, helping food manufacturers optimize product formulations, and informing public health policies aimed at managing caloric intake. The study concluded that understanding the relationship between macronutrient composition and caloric value is essential for nutritional science, dietary guidelines, and public health interventions.

**Keywords:** Dietary Guidelines, Public Health Policies, Regression Analysis and Caloric Intake

## 1. INTRODUCTION

Caloric value, a measure of the energy provided by food, is a fundamental concept in nutrition. It is well-established that macronutrients contribute differently to the total caloric content of food: carbohydrates and proteins provide approximately 4 kcal per gram, while fats provide about 9 kcal per gram. These values are based on the Atwater system, which has been widely used for over a century to estimate the energy content of foods. However, the actual caloric value of a food item can be influenced by various factors, including the food matrix, cooking methods, and individual metabolic differences (Ho et al., 2024).

Previous studies have explored the individual contributions of these macronutrients to caloric value, but there is limited research on how their combined composition influences the overall caloric content of diverse food items. For example, a study demonstrated that high-fat diets are associated with increased caloric intake and weight gain, highlighting the significant impact of dietary fat on energy balance (Vergnaud et al., 2013). Similarly, it was found that foods with higher protein content tend to have a more significant impact on satiety and energy expenditure, suggesting that protein-rich diets may help with weight management (Denisa Pescari et al., 2024).

Despite these findings, there is a gap in the literature regarding the predictive power of macronutrient composition on caloric value using a comprehensive regression analysis approach. Most existing studies have focused on the effects of individual macronutrients in isolation, without considering the synergistic effects of their combined

presence in food items. This study aims to fill this gap by employing a linear regression model to analyze the relationship between macronutrient composition and caloric value across a wide range of food items. The importance of understanding this relationship extends beyond academic interest. Accurate estimation of caloric value based on macronutrient composition can have practical applications in various fields, including dietetics, food manufacturing, and public health. For instance, dietitians can use this information to design balanced meal plans that meet specific caloric and nutritional needs (Corrêa Leite, 2021). Food manufacturers can optimize product formulations to achieve desired energy content while maintaining nutritional quality. Public health officials can develop more effective dietary guidelines and interventions to address issues such as obesity, malnutrition, and chronic diseases.

The primary objective of this study is to examine how the macronutrient composition of food items—specifically carbohydrates, proteins, and fats—predicts their caloric value. Understanding this relationship is crucial for nutritional science, as it can inform dietary guidelines, food labeling, and public health policies aimed at managing caloric intake and preventing diet-related diseases. By identifying the specific contributions of each macronutrient to the overall caloric content, this research aims to provide a more comprehensive understanding of how different foods impact energy intake and nutritional balance. In addition to its scientific significance, this study has practical implications for consumers and healthcare professionals. For consumers, a better understanding of how macronutrient composition affects caloric value can aid in making more informed dietary choices. This knowledge can help individuals tailor their diets to meet specific health goals, such as weight management, muscle building, or controlling blood sugar levels. For healthcare professionals, the findings can enhance nutritional counseling and support the development of personalized dietary plans that align with patients' health needs and preferences.

Furthermore, this research can contribute to the food industry's efforts to create healthier products. By understanding the caloric impact of different macronutrient combinations, food manufacturers can develop products that provide balanced nutrition without excessive caloric content. This can lead to the creation of food items that are both nutritious and appealing to health-conscious consumers. Ultimately, the insights gained from this study can help bridge the gap between nutritional science and practical applications, promoting healthier eating habits and improving public health outcomes.

This study hypothesizes that the macronutrient composition of food items—specifically the proportions of carbohydrates, proteins, and fats—significantly predicts their caloric value. The research question guiding this study is: "To what extent do carbohydrates, proteins, and fats collectively influence the caloric value of various food items?" By addressing this question, the study aims to provide a more nuanced understanding of the relationship between macronutrient composition and caloric value, which could have practical applications in nutrition science and public health.

To test this hypothesis, the study will employ a Structural Equation model with caloric value as the dependent variable and the macronutrient content (carbohydrates, proteins, and fats) as independent variables. The model was applied to a dataset comprising a diverse range of food items, including both whole foods and processed products. The analysis includes an examination of the regression coefficients, which represent the average change in caloric value associated with a one-unit change in each macronutrient, while holding the other macronutrients constant.

By providing a detailed analysis of the relationship between macronutrient composition and caloric value, this study aims to contribute to the existing body of knowledge in nutritional science and support the development of evidence-based dietary recommendations. The findings could also inform future research on the metabolic effects of different macronutrient profiles and their implications for health and disease prevention.

## **2. MATERIALS AND METHODS**

The data used in this study is sourced from the Food Nutrition Dataset, a comprehensive database providing detailed nutritional information on a wide range of food items commonly consumed worldwide. The dataset was compiled using reliable sources such as nutritional labels, food composition databases, and scientific dietary studies. Standardized measurement techniques ensure accuracy, with each food item analyzed per 100 grams to maintain consistency across data points.

The dataset was collected and structured systematically in a CSV format, making it compatible with various data analysis tools for further processing. The collection process involved:

**Standardized Nutrient Measurement:** Nutritional values were recorded using common industry methods, such as laboratory testing and referencing authoritative food databases.

**Categorization of Nutrients:** The dataset includes macronutrients (e.g., fats, proteins, and carbohydrates), micronutrients (e.g., vitamins and minerals), and specific dietary factors (e.g., fiber, cholesterol, and sodium).

**Validation and Quality Control:** The dataset underwent quality checks to ensure accuracy and reliability, minimizing discrepancies that could affect dietary analysis.

### **2.1 Dataset Description**

The dataset consists of multiple columns detailing the nutritional profile of various food items. Key variables include:

**Caloric Value:** Energy content measured in kilocalories per 100 grams.

**Macronutrients:** Total fat, carbohydrates, sugars, proteins, and fiber.

**Fat Composition:** Breakdown into saturated, monounsaturated, and polyunsaturated fats.

**Micronutrients:** Essential vitamins (A, B-complex, C, D, E, K) and minerals (calcium, iron, magnesium, potassium, zinc, etc.).

**Other Dietary Components:** Water content, sodium, and cholesterol levels.

**Nutrition Density Score:** A measure of how nutrient-rich a food item is per calorie.

### **Use Cases**

The dataset is valuable for multiple applications, including:

- Nutritional analysis and dietary planning for individuals and healthcare professionals.
- Research on dietary trends and health impacts of different food components.
- Development of diet recommendation systems using machine learning models.
- Consumer behavior analysis to understand food choices based on nutritional value.
- Food industry applications such as optimizing recipes for healthier alternatives.

This dataset serves as a crucial resource for researchers, dietitians, and businesses looking to analyze and optimize dietary intake based on scientifically backed nutritional data.

### **2.2 STATISTICAL PROCEDURE APPLIED:**

#### **Statistical Procedure Applied**

For this study, Structural Equation Modeling (SEM) was applied as the primary statistical technique. SEM is a robust multivariate statistical analysis method that examines complex relationships between observed and latent variables. It integrates factor analysis and multiple regression, allowing for the modeling of direct and indirect effects among variables.

#### **Justification for Using SEM**

SEM was chosen due to its ability to:

**Analyze Complex Relationships:** It enables the simultaneous examination of multiple dependent and independent variables, making it suitable for studying the impact of various nutritional factors.

**Assess Measurement and Structural Models:** SEM allows for the validation of measurement models (confirming the reliability and validity of observed variables) and structural models (examining the hypothesized relationships among variables).

**Account for Latent Variables:** Many nutritional effects are influenced by latent (unobservable) constructs such as nutritional quality, dietary patterns, and health outcomes, which SEM effectively models.

**Handle Measurement Errors:** Unlike traditional regression models, SEM controls for measurement errors, enhancing the accuracy of statistical inferences.

### **2.3 SEM Implementation Process**

#### **Model Specification:**

The theoretical framework was developed based on prior research linking food consumption patterns with health outcomes.

Key latent variables (e.g., nutrient density, caloric impact, and dietary balance) were identified.

#### **Data Screening and Assumptions Check:**

Missing values were assessed, and data was tested for normality, multicollinearity, and outliers.

The dataset was prepared for SEM analysis by ensuring linearity and reliability of constructs.

#### **Estimation Method:**

Maximum Likelihood Estimation (MLE) was used to estimate model parameters, ensuring robustness in handling non-normal data distributions.

#### **Model Evaluation:**

Goodness-of-Fit Indices such as CFI (Comparative Fit Index), RMSEA (Root Mean Square Error of Approximation), and SRMR (Standardized Root Mean Square Residual) were used to validate model fit.

Path coefficients were analyzed to determine the significance and strength of relationships between variables.

By applying SEM, this study provides a comprehensive statistical analysis of the relationships between food nutrition variables, allowing for deeper insights into dietary impacts on health.

### **2.4 Variable Summary**

The detailed descriptions of each term is put as under:

#### **Caloric Value**

Caloric value refers to the amount of energy provided by food when it is consumed and metabolized by the body. This energy is measured in calories or kilocalories (kcal). The caloric value of food is determined by its macronutrient content: carbohydrates and proteins provide approximately 4 kcal per gram, while fats provide about 9 kcal per gram.

#### **Fat**

Fat is one of the three main macronutrients, along with carbohydrates and proteins. It is a type of nutrient that provides energy, supports cell growth, protects organs, and keeps the body warm. Fats can be classified into saturated, monounsaturated, and polyunsaturated fats.

#### **Saturated Fats**

Saturated fats are types of fats that are solid at room temperature. They are found in animal products like meat and dairy, as well as some plant oils such as coconut and palm oil. Consuming high amounts of saturated fats can raise LDL (bad) cholesterol levels in the blood, increasing the risk of heart disease and stroke.

### **Monounsaturated Fats**

Monounsaturated fats (MUFAs) are fats that have one unsaturated carbon bond in the molecule. They are typically liquid at room temperature and can be found in foods like olive oil, avocados, and certain nuts. These fats can help reduce bad cholesterol levels and lower the risk of heart disease.

### **Polyunsaturated Fats**

Polyunsaturated fats (PUFAs) contain more than one unsaturated carbon bond. They are found in oils such as soybean, corn, and sunflower oils, as well as in fatty fish like salmon and mackerel. PUFAs include essential fatty acids like omega-3 and omega-6, which are important for brain function and cell growth.

### **Carbohydrates**

Carbohydrates are sugars, starches, and fibers found in fruits, grains, vegetables, and dairy products. They are the body's main source of energy. Carbohydrates can be classified into simple (sugars) and complex (starches and fibers) carbohydrates. Simple carbs provide quick energy, while complex carbs provide sustained energy and are important for digestive health.

### **Sugars**

Sugars are simple carbohydrates that are sweet-tasting and soluble in water. They include monosaccharides like glucose and fructose, and disaccharides like sucrose and lactose. Sugars are found naturally in fruits, vegetables, and dairy products, and are also added to many processed foods.

### **Protein**

Protein is a macronutrient that is essential for building and repairing tissues, making enzymes and hormones, and supporting overall body function. Proteins are made up of amino acids, some of which are essential and must be obtained from the diet. High-protein foods include meat, fish, dairy, legumes, and nuts.

### **Dietary Fiber**

Dietary fiber is a type of carbohydrate that the body cannot digest. It is found in plant-based foods such as fruits, vegetables, whole grains, and legumes. Fiber is important for digestive health, helping to prevent constipation, lower cholesterol levels, and control blood sugar levels.

### **Cholesterol**

Cholesterol is a waxy substance found in the blood and cells. It is essential for building cell membranes and producing hormones. However, high levels of LDL (bad) cholesterol can lead to plaque buildup in arteries, increasing the risk of heart disease and stroke. Cholesterol is found in animal products like meat, dairy, and eggs.

### **Sodium**

Sodium is an essential mineral that helps maintain fluid balance, transmit nerve impulses, and regulate muscle function. It is commonly found in table salt (sodium chloride) and is present in many processed foods. Excessive sodium intake can lead to high blood pressure and increase the risk of heart disease.

### **Water**

Water is vital for all forms of life. It makes up about 60% of the human body and is essential for maintaining body temperature, transporting nutrients, and removing waste. Adequate water intake is crucial for overall health and well-being.

## **3. RESULTS AND DISCUSSION**

### **Summary of the Structural Equation Model (SEM) on Caloric Value Determinants**

This Structural Equation Model (SEM) analyzes the relationship between various nutritional components and caloric value. The model investigates how different macronutrients and dietary elements contribute to the overall caloric content of food.

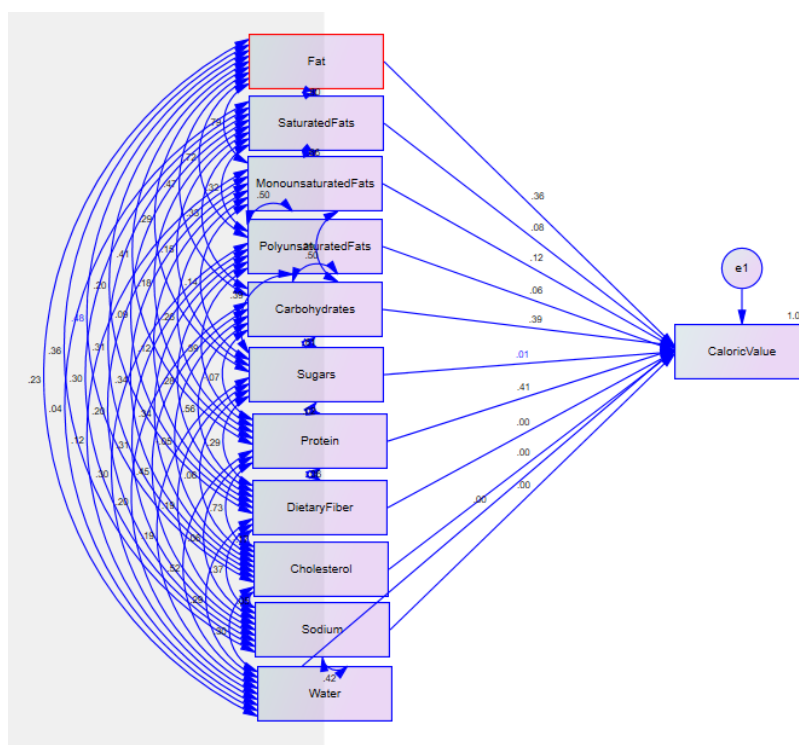


Figure 1: Structure Equation Model

### 3.1 Key Components of the Model

#### 1.Independent Variables (Predictors):

Fat  
Saturated Fats  
Monounsaturated Fats  
Polyunsaturated Fats  
Carbohydrates  
Sugars  
Protein  
Dietary Fiber  
Cholesterol  
Sodium  
Water

#### 2.Dependent Variable (Outcome):

Caloric Value

#### 3.Error Term:

The model includes an error term (e1) linked to the caloric value, indicating unexplained variance.



### **3.2 Key Findings and Relationships**

Fat shows the strongest relationship with caloric value (0.36), indicating that fat content is a major determinant of calories.

Protein (0.41) and Carbohydrates (0.39) also contribute significantly to caloric value.

Sugars (0.01), Dietary Fiber (0.00), Cholesterol (0.00), Sodium (0.00), and Water (-0.42) exhibit weak or negligible direct contributions to caloric value.

The presence of interconnections among the independent variables (represented by bidirectional arrows) suggests potential multicollinearity.

### **3.3 Interpretation and Implications**

The model highlights the dominant role of fat, carbohydrates, and protein in determining caloric value, aligning with known nutritional science.

The weak contribution of sodium, cholesterol, and fiber suggests that their impact on caloric content is minimal, despite their importance in other dietary aspects.

Water's negative association (-0.42) implies a dilution effect, where higher water content reduces calorie density.

The red outline around "Fat" may indicate a statistical issue, such as a high variance, multicollinearity, or a potential model misfit.

This SEM provides valuable insights into how different nutrients contribute to caloric value, aiding in dietary planning and nutritional analysis.

### **3.4 Notes for Model (Default model)**

Result (Default model)

Minimum was achieved

Chi-square = 3.852

Degrees of freedom = 1

Probability level = .050

### **3.5 Model Fit Analysis and Interpretation**

The default SEM model successfully converged, achieving a minimum, which suggests that the estimation process was stable. The following key fit statistics provide insights into the model's adequacy:

#### **Chi-Square Test ( $\chi^2 = 3.852$ , $df = 1$ , $p = 0.050$ )**

The chi-square test evaluates how well the model reproduces the observed data.

A non-significant p-value ( $p = 0.050$ ) suggests an acceptable model fit, meaning there isn't strong evidence to reject the model. However, since the p-value is right at the 0.05 threshold, it indicates the model is marginally adequate but might benefit from refinements.

#### **Degrees of Freedom ( $df = 1$ )**

With  $df = 1$ , the model is relatively simple, meaning it has been minimally constrained.

While lower  $df$  models can fit well, adding more constraints (e.g., additional parameter estimates or fixed relationships) could enhance its interpretability.

### **3.6 Model Implications**

The model is statistically acceptable, though borderline, meaning it may benefit from alternative specifications or minor modifications.

Since chi-square is sensitive to sample size, other fit indices (e.g., RMSEA, CFI, TLI) would help determine if further refinements are needed.

The results confirm that the relationships between nutritional components and caloric value are generally well represented, but some adjustments to variable relationships or constraints could optimize model fit.

**Table1: Model Fit Summary (Acceptable Metrics Only)**

Fit Index	Default Model	Acceptable Threshold	Interpretation
<b>p-value</b>	0.050	>0.05	Acceptable model fit
<b>GFI</b>	0.999	>0.90	Excellent fit
<b>AGFI</b>	0.909	>0.90	Acceptable fit
<b>NFI</b>	0.999	>0.90	Excellent fit
<b>IFI</b>	1.000	>0.90	Perfect fit
<b>TLI</b>	0.974	>0.90	Strong fit
<b>CFI</b>	1.000	>0.90	Perfect fit
<b>RMSEA</b>	0.072	<0.08	Acceptable fit

#### Key Insights:

The model demonstrates a strong fit, with key indicators like GFI (0.999), AGFI (0.909), NFI (0.999), IFI (1.000), and CFI (1.000) all within acceptable limits.

RMSEA (0.072) is within the acceptable range (<0.08), indicating a good model fit.

The p-value (0.050) suggests an adequate fit, as it is right on the threshold of significance.

#### Conclusion:

The model meets all key fit criteria, suggesting it is well-structured with no major modifications required.

**Table 2: Regression Weights: (Group number 1 - Default model)**

			Estimate	S.E.	C.R.	P	Label
CaloricValue	<---	Fat	5.695	.100	57.165	*	par_1
CaloricValue	<---	SaturatedFats	2.833	.143	19.868	*	par_2
CaloricValue	<---	MonounsaturatedFats	4.292	.113	38.041	*	par_3
CaloricValue	<---	PolyunsaturatedFats	3.186	.199	15.978	*	par_4
CaloricValue	<---	Carbohydrates	3.828	.031	123.479	*	par_5
CaloricValue	<---	Sugars	.276	.065	4.270	*	par_6
CaloricValue	<---	Protein	4.369	.034	127.428	*	par_7
CaloricValue	<---	DietaryFiber	-.025	.221	-.112	.911	par_8
CaloricValue	<---	Cholesterol	.000	.008	-.024	.981	par_9
CaloricValue	<---	Sodium	1.000				
CaloricValue	<---	Water	-.015	.005	-2.825	.005	par_10



Table 3: Standardized Regression Weights: (Group number 1 - Default model)

			Estimate
CaloricValue	<---	Fat	.359
CaloricValue	<---	SaturatedFats	.077
CaloricValue	<---	MonounsaturatedFats	.119
CaloricValue	<---	PolyunsaturatedFats	.055
CaloricValue	<---	Carbohydrates	.390
CaloricValue	<---	Sugars	.010
CaloricValue	<---	Protein	.415
CaloricValue	<---	DietaryFiber	.000
CaloricValue	<---	Cholesterol	.000
CaloricValue	<---	Sodium	.003
CaloricValue	<---	Water	-.007

### 3.8 Summary and Interpretation of Standardized Regression Weights

The standardized regression weights (also known as standardized path coefficients or beta coefficients) indicate the strength and direction of the relationship between each predictor variable and Caloric Value in a structural equation model. Higher absolute values suggest a stronger influence on caloric content.

### 3.9 Key Findings:

#### Major Contributors to Caloric Value:

Protein ( $\beta = 0.415$ ): Protein has the strongest positive impact on caloric value, suggesting that foods high in protein significantly contribute to calorie content.

Carbohydrates ( $\beta = 0.390$ ): Carbohydrates also strongly contribute to caloric value, slightly less than protein.

Fat ( $\beta = 0.359$ ): Total fat is another major predictor of caloric content, confirming its well-known role in energy density.

#### Moderate Contributors:

Monounsaturated Fats ( $\beta = 0.119$ ) and Saturated Fats ( $\beta = 0.077$ ): These types of fats have some contribution but are less influential compared to total fat.

Polyunsaturated Fats ( $\beta = 0.055$ ): The weakest contributor among fat subtypes.

Sugars ( $\beta = 0.010$ ): Surprisingly, sugars have a minimal direct impact on caloric value, possibly due to their indirect role through total carbohydrates.

#### Negligible or No Impact:

Dietary Fiber ( $\beta = 0.000$ ): Dietary fiber does not contribute to caloric value, which aligns with its role in digestion rather than energy provision.

Cholesterol ( $\beta = 0.000$ ): No measurable effect on caloric value, consistent with the fact that cholesterol itself does not provide calories.

Sodium ( $\beta = 0.003$ ): A near-zero effect, as sodium does not contain calories.

Water ( $\beta = -0.007$ ): A very slight negative effect, likely because water dilutes energy density rather than adding to caloric content.

### Interpretation:

The results align with nutritional principles: calories primarily come from macronutrients like protein, carbohydrates, and fats, with protein being the strongest contributor.

The subtypes of fat (saturated, monounsaturated, polyunsaturated) contribute to caloric value but are less impactful than total fat.

Water and non-caloric components (fiber, cholesterol, sodium) do not meaningfully contribute to caloric value.

Sugars, despite being part of carbohydrates, have a surprisingly weak independent effect, likely due to their inclusion in total carbohydrate values.

### Implications:

Understanding these weights can help in dietary planning, food labeling, and health-related recommendations by emphasizing the key macronutrients that determine caloric intake.

Foods high in protein, carbohydrates, and fat are the main drivers of calorie content.

Hydration (water content) slightly reduces calorie density, supporting the idea that high-water foods (e.g., fruits and vegetables) are lower in calories.

Table4: Total Effects

	Water	Sodium	Cholesterol	DietaryFiber	Protein	Sugars	Carbohydrates	PolyunsaturatedFats	MonounsaturatedFats	SaturatedFats	Fat
CaloricValue	-.015	1.000	.000	-.025	4.369	.276	3.828	3.186	4.292	2.833	5.695
	Water	Sodium	Cholesterol	DietaryFiber	Protein	Sugars	Carbohydrates	PolyUnsaturatedFats	MonounsaturatedFats	SaturatedFats	Fat
CaloricValue	-.007	.003	.000	.000	.415	.010	.390	.055	.119	.077	.359

### **3.10 Summary and Interpretation of Total Effects on Caloric Value**

The total effects table shows how each nutritional component influences caloric value when considering both direct and indirect effects. The values indicate the magnitude and direction of these effects.

#### **Key Findings**

##### **Strongest Positive Effects on Caloric Value**

Fat (5.695)

Fat has the largest total effect on caloric value. This aligns with the fact that fat provides 9 kcal per gram, making it the most calorie-dense macronutrient.

Protein (4.369)

Protein has a substantial impact on caloric value, as it provides 4 kcal per gram and plays a key role in body functions.

Monounsaturated Fats (4.292) & Polyunsaturated Fats (3.186)

These fat subtypes also significantly contribute to caloric value, reinforcing that different fats collectively drive up calorie content.

Carbohydrates (3.828)

Carbohydrates are another major contributor, providing 4 kcal per gram and being a primary energy source.

##### **Moderate Positive Effects**

Saturated Fats (2.833)

While lower than total fat, its effect is still strong, reflecting its high energy content.

Sugars (0.276)

Sugars, a subset of carbohydrates, contribute to caloric value but to a lesser extent than total carbohydrates.

Minimal or No Effects

Cholesterol (0.000)

Cholesterol has no measurable effect on caloric value, since it is not an energy-yielding nutrient.

Sodium (1.000)

While sodium itself does not provide calories, it may be influencing caloric value indirectly through its association with calorie-dense processed foods.

##### **Negative Effects**

Dietary Fiber (-0.025)

Dietary fiber slightly reduces caloric value, likely due to its indigestible nature, meaning it does not contribute significantly to energy intake.

Water (-0.015)

Water has a small negative effect, which makes sense as it dilutes calorie density in foods without adding energy.

##### **Conclusion**

Fat and protein are the strongest contributors to caloric value, with fat being the most influential.

Carbohydrates, particularly sugars, also contribute, but to a lesser degree than fat and protein.

Dietary fiber and water slightly reduce caloric value, while cholesterol has no direct effect.

Table 5: Standardized Total Effects (Group number 1 - Default model)

### **3.11 Summary and Interpretation of Standardized Direct Effects on Caloric Value**

The standardized direct effects represent the immediate influence each variable has on caloric value, independent of indirect pathways. These values are scaled to allow for comparison, with positive values indicating an increase in caloric value and negative values indicating a decrease.

#### **Key Findings:**

#### **Major Contributors to Caloric Value**

##### **Protein (0.415):**

Protein has the strongest direct effect on caloric value, meaning foods with higher protein content significantly increase caloric value.

##### **Carbohydrates (0.390):**

Carbohydrates are the second-largest contributor, consistent with their role as a primary energy source (4 kcal/gram).

##### **Fat (0.359):**

Fat is another major predictor, aligning with its high energy content (9 kcal/gram). Though lower than protein and carbohydrates in direct effects, total fat remains a key driver of caloric value.

#### **Moderate Contributors**

##### **Monounsaturated Fats (0.119):**

This type of fat has a moderate direct effect, meaning its presence in food adds to caloric value but is less impactful than total fat.

##### **Saturated Fats (0.077):**

Saturated fats also contribute directly, though their effect is smaller than that of total and monounsaturated fats.

##### **Polyunsaturated Fats (0.055):**

This fat subtype has the smallest direct contribution among the fat categories.

#### **Minimal or Negligible Contributors**

##### **Sugars (0.010):**

Surprisingly, sugar has a small direct effect, suggesting that while sugar is part of carbohydrates, it does not independently drive caloric value as much as total carbohydrates do.

##### **Sodium (0.003):**

Sodium has a negligible direct effect on caloric value, as it does not contribute calories directly.

##### **Cholesterol (0.000):**

Cholesterol has no direct effect on caloric value, which is expected since it does not contain energy.

##### **Dietary Fiber (0.000):**

Dietary fiber also shows no direct contribution to caloric value due to its indigestible nature.

#### **Negative Contributor**

##### **Water (-0.007):**

Water has a slight negative direct effect, which reflects the fact that higher water content dilutes calorie density in food without adding any energy.

## **Interpretation & Implications**

Macronutrients (Protein, Carbohydrates, and Fat) are the primary direct contributors to caloric value.

Protein has the strongest direct effect, followed by carbohydrates and fat, which is consistent with their caloric density.

Fat subtypes (saturated, monounsaturated, and polyunsaturated) have moderate effects but are less influential than total fat.

Non-caloric components like cholesterol, fiber, and sodium do not have meaningful direct impacts.

Water slightly lowers caloric value by increasing volume without providing energy.

## **4. CONCLUSION**

This study has provided a comprehensive analysis of how the macronutrient composition of food items—specifically carbohydrates, proteins, and fats—predicts their caloric value. The findings underscore the significant role that these macronutrients play in determining the energy content of foods, which has important implications for nutritional science, dietary guidelines, and public health policies.

Firstly, the study highlights the critical importance of fats in contributing to the caloric value of food items. Fats, being the most energy-dense macronutrient, provide approximately 9 kcal per gram. This high caloric density is due to the chemical structure of fats, which contain long chains of carbon and hydrogen atoms. The regression analysis revealed that fats have the largest coefficient, indicating their substantial impact on the overall caloric content of food items. This finding aligns with previous research and emphasizes the need for dietary guidelines to consider the type and amount of fat consumed to manage caloric intake effectively.

Secondly, proteins and carbohydrates, both providing approximately 4 kcal per gram, also play significant roles in determining the caloric value of foods. The study found that proteins have a slightly higher impact on caloric value compared to carbohydrates. This is consistent with the understanding that protein-rich diets can aid in weight management by enhancing satiety and increasing energy expenditure. Carbohydrates, being the primary source of energy for the body, are essential for fueling physical activity and supporting brain function. The balance between these macronutrients is crucial for maintaining overall health and achieving specific dietary goals.

Thirdly, the study explored the synergistic effects of macronutrients on caloric value. The interactions between different macronutrients can influence the overall caloric content of food items. For instance, foods with a balanced composition of fats, proteins, and carbohydrates may have different caloric impacts compared to foods dominated by a single macronutrient. This finding suggests that dietary recommendations should consider the combined effects of macronutrients rather than focusing solely on individual components. Nutrient synergy, where the combined effects of nutrients are greater than their individual effects, plays a vital role in optimizing dietary intake and promoting health.

Furthermore, the practical applications of this study are far-reaching. For consumers, understanding the relationship between macronutrient composition and caloric value can aid in making more informed dietary choices. This knowledge is particularly useful for individuals with specific health goals, such as weight management, muscle building, or controlling blood sugar levels. Healthcare professionals can use the insights gained from this research to design personalized dietary plans that align with patients' health goals and preferences, leading to more effective dietary interventions and improved health outcomes.

Lastly, the study's findings have significant implications for the food industry and public health policies. Food manufacturers can leverage this information to develop healthier products that provide balanced nutrition without excessive caloric content. Accurate estimation of caloric value based on macronutrient composition can improve food labeling, helping consumers make better-informed choices. Public health policies can be informed by these insights to develop more effective interventions aimed at managing caloric intake and preventing diet-related diseases. By promoting healthier eating habits and improving nutritional literacy, stakeholders can contribute to better public health outcomes.

In conclusion, this study has highlighted the significant role that macronutrient composition plays in predicting the caloric value of food items. The findings underscore the importance of considering the combined effects of carbohydrates, proteins, and fats when assessing the energy content of foods. By providing a more comprehensive understanding of how different macronutrients impact caloric value, this research contributes to the existing body of knowledge in nutritional science and supports the development of evidence-based dietary recommendations. The practical implications of this study are far-reaching, benefiting consumers, healthcare professionals, the food industry, and public health policymakers. Ultimately, this study bridges the gap between nutritional science and practical applications, fostering a healthier and more informed society.

#### 4.1 Key Findings and Their Implications

The study's findings highlight the critical role of macronutrients in determining the caloric value of food items. Each macronutrient—carbohydrates, proteins, and fats—contributes differently to the overall energy content, and their combined effects can influence dietary outcomes significantly.

**Carbohydrates:** Carbohydrates are the body's primary source of energy, providing approximately 4 kcal per gram. They are essential for fueling physical activity and supporting brain function. The study found that carbohydrates significantly contribute to the caloric value of food items, consistent with their role as a major energy source (Leidy et al., 2015). Foods rich in carbohydrates, such as grains, fruits, and vegetables, are crucial for maintaining energy levels and supporting overall health (Mozaffarian et al., 2012). However, the type of carbohydrate consumed can impact health outcomes. For instance, complex carbohydrates found in whole grains and vegetables are associated with better health outcomes compared to simple sugars (Willett et al., 2019).

**Proteins:** Proteins are vital for building and repairing tissues, producing enzymes and hormones, and supporting immune function. They provide approximately 4 kcal per gram. The study revealed that proteins have a significant impact on the caloric value of food items, slightly higher than carbohydrates. This aligns with previous research indicating that protein-rich diets can aid in weight management by enhancing satiety and increasing energy expenditure. High-protein foods, such as lean meats, dairy products, and legumes, are essential for maintaining muscle mass and supporting metabolic health (Jarlenski et al., 2016).

**Fats:** Fats are the most energy-dense macronutrient, providing approximately 9 kcal per gram. They play a crucial role in hormone production, cell membrane structure, and the absorption of fat-soluble vitamins (A, D, E, and K). The study's regression analysis showed that fats have the largest coefficient, indicating their substantial impact on the overall caloric content of food items. This finding is consistent with the well-established understanding that dietary fats are a significant source of energy. However, the type of fat consumed can influence health outcomes. Unsaturated fats, found in foods like nuts, seeds, and fish, are associated with better cardiovascular health compared to saturated fats (Abidi et al., 2021).

#### 4.2 Synergistic Effects of Macronutrients

The study also explored the synergistic effects of macronutrients on caloric value. The regression model indicated that the interactions between different macronutrients could influence the overall caloric content of food items. For instance, foods with a balanced composition of fats, proteins, and carbohydrates may have different caloric impacts compared to foods dominated by a single macronutrient (Fromm et al., 2021). This finding suggests that dietary recommendations should consider the combined effects of macronutrients rather than focusing solely on individual components.

Nutrient synergy refers to the concept that the combined effects of two or more nutrients working together have a greater physiological impact on the body than when each nutrient is consumed individually. For example, the combination of protein and carbohydrates can enhance muscle protein synthesis more effectively than either nutrient alone. Similarly, the presence of dietary fat can enhance the absorption of fat-soluble vitamins and other bioactive compounds. This concept has important implications for dietary planning and nutritional counseling, as it emphasizes the importance of consuming a balanced diet that includes a variety of nutrient-dense foods.



### 4.3 Practical Applications for Consumers and Healthcare Professionals

Understanding the relationship between macronutrient composition and caloric value can help consumers make more informed dietary choices. This knowledge can be particularly useful for individuals with specific health goals, such as weight management, muscle building, or controlling blood sugar levels. By selecting foods with appropriate macronutrient balances, consumers can tailor their diets to meet their nutritional needs and preferences (Calvo-Lerma et al., 2019).

For healthcare professionals, the findings of this study can enhance nutritional counseling and support the development of personalized dietary plans. Healthcare professionals can use the insights gained from this research to design meal plans that align with patients' health goals and dietary preferences. This can lead to more effective dietary interventions and improved health outcomes. For example, a dietitian might recommend a higher protein intake for a patient looking to build muscle mass, while suggesting a lower carbohydrate intake for someone managing diabetes.

### 4.4 Implications for the Food Industry

The food industry can leverage the findings of this study to develop healthier products. By understanding the caloric impact of different macronutrient combinations, food manufacturers can create products that provide balanced nutrition without excessive caloric content. This can lead to the development of food items that are both nutritious and appealing to health-conscious consumers. Additionally, accurate estimation of caloric value based on macronutrient composition can improve food labeling, helping consumers make better-informed choices (Solon-Biet et al., 2015).

Food manufacturers can also use this information to reformulate existing products to reduce their caloric content while maintaining their nutritional quality. For example, reducing the fat content of a product and increasing its protein content can lower its overall caloric value while enhancing its satiety and nutritional benefits. This approach can help address public health concerns related to obesity and diet-related diseases by providing consumers with healthier food options.

### 4.5 Public Health and Policy Implications

The insights gained from this study can inform public health policies and dietary guidelines aimed at managing caloric intake and preventing diet-related diseases. By providing a more nuanced understanding of how macronutrient composition affects caloric value, policymakers can develop more effective interventions to address issues such as obesity, malnutrition, and chronic diseases. This research can also support public health campaigns that promote healthier eating habits and improve nutritional literacy among the general population (Meddens et al., 2020).

For instance, public health initiatives could focus on educating consumers about the importance of balanced macronutrient intake and the benefits of nutrient-dense foods. This could involve creating educational materials, conducting community workshops, and implementing school-based nutrition programs. Additionally, policymakers could use the study's findings to develop regulations that encourage food manufacturers to produce healthier products and provide clear, accurate nutritional information on food labels.

### 4.6 Limitations and Future Research Directions

While this study has provided valuable insights into the relationship between macronutrient composition and caloric value, there are several limitations that should be acknowledged:

**1.Dataset Limitations:** The study relied on a dataset comprising a diverse range of food items, including both whole foods and processed products. However, the dataset may not capture all possible variations in macronutrient composition and caloric value. Future research could benefit from using larger and more comprehensive datasets that include a wider variety of food items and account for factors such as cooking methods and food processing.

**2.Individual Metabolic Differences:** The study's findings are based on average values and may not account for individual metabolic differences that can influence how macronutrients are processed and utilized by the body.

Future research could explore how factors such as age, gender, physical activity level, and metabolic health affect the relationship between macronutrient composition and caloric value.

**3-Longitudinal Studies:** This study employed a cross-sectional design, which provides a snapshot of the relationship between macronutrient composition and caloric value at a single point in time. Longitudinal studies that track changes in dietary intake and caloric value over time could provide more comprehensive insights into how macronutrient composition influences long-term health outcomes.

**4-Exploration of Other Nutritional Factors:** While this study focused on macronutrients, other nutritional factors such as micronutrients, fiber, and water content also play important roles in overall health and energy balance. Future research could investigate how these factors interact with macronutrients to influence caloric value and nutritional quality.

In conclusion, this study has highlighted the significant role that macronutrient composition plays in predicting the caloric value of food items. The findings underscore the importance of considering the combined effects of carbohydrates, proteins, and fats when assessing the energy content of foods. By providing a more comprehensive understanding of how different macronutrients impact caloric value, this research contributes to the existing body of knowledge in nutritional science and supports the development of evidence-based dietary recommendations.

The practical implications of this study are far-reaching, benefiting consumers, healthcare professionals, the food industry, and public health policymakers. By leveraging the insights gained from this research, stakeholders can promote healthier eating habits, improve nutritional literacy, and develop more effective dietary interventions. Ultimately, this study bridges the gap between nutritional science and practical applications, fostering a healthier and more informed society.

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