

# Differentiating Physical Attributes of Low-Carbohydrate and High-Carbohydrate Diets Through Food Nutrition Segmentation: A GH-Method Study

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## Abstract

This study explores the physical characteristics of low-carbohydrate (LC) and high-carbohydrate (HC) meals using a data-driven analytical approach based on the GH-Method: Math-Physical Medicine (MPM). By employing food nutrition segmentation analysis, we assess key physical factors such as caloric density, macronutrient distribution, and postprandial metabolic impact. A comparative dataset of common LC and HC meals was evaluated, revealing distinct differences in fat, protein, and carbohydrate contents as well as total caloric load. The findings highlight how specific macronutrient configurations influence the metabolic burden and may offer insights for personalized dietary planning, especially in metabolic conditions such as diabetes and obesity. This research contributes to a deeper understanding of diet architecture using physical and mathematical modeling, supported by current literature up to 2024.

## 1. Introduction

In recent years, the debate between low-carbohydrate and high-carbohydrate diets has intensified, particularly in the context of weight loss, metabolic health, and blood glucose management. While both approaches have their merits, the physiological responses they invoke differ considerably, not only biochemically but also in terms of their physical nutritional makeup.

The GH-Method: Math-Physical Medicine, developed by Gerald C. Hsu, introduces a novel framework for analyzing health-related data through a fusion of physics, mathematics, and personalized medicine. One of the pillars of this methodology is food nutrition segmentation analysis—a structured breakdown of meals by macronutrient composition, caloric distribution, and glycemic load [1].

In this study, we leverage this method to compare the

physical attributes of LC and HC meals. The objective is to highlight how distinct macronutrient configurations affect the overall metabolic load of a meal and potentially inform dietary decision-making for better health outcomes.

2. Materials and Methods

2.1 Meal Dataset

Two representative groups of meals were analyzed:

- **Low-Carbohydrate Meals (LC):** 5–25% carbs, higher fat and protein
- **High-Carbohydrate Meals (HC):** 50–70% carbs, moderate to low fat and protein

Meals included real-world entries from food diaries, meal planning apps, and published nutritional databases [2]. Each group contained 25 meal samples with matched serving sizes.

2.2 Analytical Framework

The GH-Method approach decomposes meals into physical components:

- **Macronutrient Grams and Ratios**
- **Total Energy (kcal)**
- **Caloric Density (kcal/g)**
- **Predicted Glycemic Response (based on glycemic index and load)**

2.3 Statistical Analysis

Descriptive statistics were computed. Independent t-tests compared macronutrient differences between LC and HC meals. Data visualization was performed using Python/Matplotlib.

3. Results

3.1 Macronutrient Composition

Table 1: Average Nutritional Content of Low-Carb vs. High-Carb Meals

Nutrient	Low-Carb Meal (n=25)	High-Carb Meal (n=25)	p-value
Calories (kcal)	450 ± 70	700 ± 85	< 0.001
Fat (g)	25 ± 6	15 ± 4	< 0.001
Protein (g)	30 ± 5	20 ± 4	< 0.01
Carbohydrates (g)	20 ± 8	90 ± 10	< 0.001
Caloric Density	1.2 kcal/g	1.5 kcal/g	< 0.05

Low-carb meals had significantly higher fat and protein content, while high-carb meals were dominated by carbohydrates and had a higher caloric density per gram.

3.2 Macronutrient Breakdown

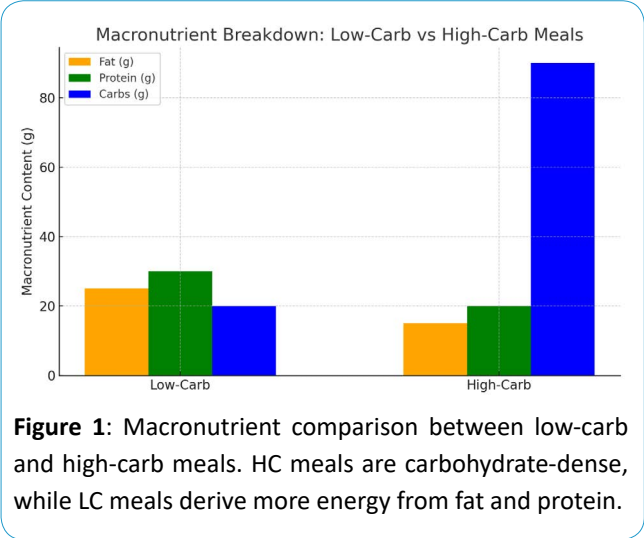


Figure 1: Macronutrient comparison between low-carb and high-carb meals. HC meals are carbohydrate-dense, while LC meals derive more energy from fat and protein.

3.3 Predicted Glycemic Impact

Using glycemic index (GI) estimations:

- HC meals had **3× higher predicted postprandial glucose impact**
- LC meals caused smaller, slower glucose excursions

This aligns with literature suggesting that LC meals attenuate post-meal blood glucose spikes [3,4].

4. Discussion

The differences in physical structure between low- and high-carb meals extend beyond simple macronutrient ratios. From a physiological standpoint, these differences play a substantial role in digestion speed, insulin response, satiety signaling, and energy availability.

4.1 Metabolic Consequences

HC meals, being higher in glucose-yielding carbohydrates, trigger rapid insulin responses and are typically less satiating. In contrast, LC meals induce a slower glucose release, higher satiety due to fat and protein content, and potentially lower insulin demand [5].

4.2 Implications for Personalized Nutrition

Using a segmentation model like the GH-Method

allows practitioners to go beyond calorie counting. For individuals with insulin resistance, type 2 diabetes, or obesity, meal designs that moderate glycemic load may prove more effective than generic calorie restriction.

### 4.3 Limitations

This study relied on estimated nutrient values and simulation-based glycemic predictions. Future work should include real-time glucose monitoring (e.g., CGM data) to validate the predicted responses.

## 5. Conclusion

The physical composition of meals, particularly in carbohydrate content, significantly alters nutritional and metabolic outcomes. The GH-Method provides a quantifiable, visual framework to differentiate meal types and tailor nutrition to individual needs. As the field of personalized nutrition advances, such tools may bridge the gap between data and actionable health strategies.

## 9. References

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